

RATIOS OF PARTITION FUNCTIONS FOR THE LOG-GAMMA POLYMER

NICOS GEORGIOU, FIRAS RASSOUL-AGHA, TIMO SEPPÄLÄINEN, AND ATILLA YILMAZ

ABSTRACT. We introduce a random walk in random environment associated to an underlying directed polymer model in 1+1 dimensions. This walk is the positive temperature counterpart of the competition interface of percolation and arises as the limit of quenched polymer measures. We prove this limit for the exactly solvable log-gamma polymer, as a consequence of almost sure limits of ratios of partition functions. These limits of ratios give the Busemann functions of the log-gamma polymer, and furnish correctors that solve a variational formula for the limiting free energy. Limits of ratios of point-to-point and point-to-line partition functions manifest a duality between tilt and velocity that comes from quenched large deviations under polymer measures. In the log-gamma case we identify a family of ergodic invariant distributions for the random walk in random environment.

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1. INTRODUCTION

In directed polymer models the definition of weak disorder is that normalized point-to-line partition functions converge to a strictly positive random variable. In strong disorder these normalized partition functions converge to zero. Weak disorder takes place only in dimensions $3+1$ and higher and under high enough temperature, while lower dimensions are in strong disorder throughout the temperature range. (See [3, 5, 6, 7, 8, 12, 20] for reviews and some key results.)

We work in $1+1$ dimensions with the explicitly solvable log-gamma polymer. We show that ratios of both point-to-point partition functions and tilted point-to-line partition functions converge almost surely to gamma-distributed limits. Out of this basic fact we derive several consequences.

(i) Limits of ratios of partition functions give us limits of quenched polymer measures, both point-to-point and point-to-line, as the path length tends to infinity. The limit processes can be regarded as infinitely long polymers. Technically they are random walks in correlated random environments (RWRE). When we average over the environment, this RWRE has fluctuation exponent $2/3$, in accordance with $1+1$ dimensional Kardar-Parisi-Zhang universality. This polymer RWRE is also a positive temperature counterpart of a competition interface in a percolation model. (This terminology comes from the idea that percolation models are zero-temperature polymers. Remark 2.1 below explains.) For the RWRE we identify a family of stationary and ergodic distributions for the environment as seen from the particle. The averaged stationary RWRE is a standard random walk.

(ii) Logarithms of the limiting point-to-point ratios give us an analogue of Busemann functions in the positive temperature setting. Busemann functions have emerged as a central object in the study of geodesics and invariant distributions of percolation models and related interacting particle systems [1, 4, 11, 16, 21]. Our paper introduces this notion in the positive temperature setting. We show how Busemann functions solve a variational problem that characterizes the limiting free energy density of the log-gamma polymer.

A theme that appears more than once is a familiar large deviations duality between the asymptotic velocity of the path under polymer distributions and a tilt introduced into the partition function and probability distribution. In this duality the mapping from velocity to tilt is given by the expectation of the Busemann function. In particular, this duality determines how limits of ratios of point-to-point and tilted point-to-line partition functions match up with each other.

A word of explanation about our focus on the log-gamma polymer. The ultimate goal is of course to find results valid for a wide class of polymer models. We could formulate at least some of our results more generally. But the statements would be complicated and need hypotheses that we can presently verify only for the log-gamma model anyway. For general polymers, just as for general percolation models, we cannot currently prove even mild regularity properties for the limiting free energy. Thus we chose to focus exclusively on the log-gamma model (except for the general discussions in Sections 2 and 5).

We expect that much of this picture can eventually be verified for general $1+1$ dimensional directed polymers. Our hope is that this paper would inspire such further work. For example, it is clear that the solution of the variational formula for the free energy in terms of

Busemann functions works completely generally, once a sufficiently strong existence statement for Busemann functions is proved. Busemann functions with tractable distributions are an essential feature of the exact solvability of the log-gamma polymer. They can be used to construct a shift-invariant version of the polymer model, which was earlier used for deriving fluctuation exponents and large deviation rate functions [15, 27].

The log-gamma polymer is a canonical model in the Kardar-Parisi-Zhang universality class, in the same vein as the asymmetric simple exclusion process and the corner growth model with geometric or exponential weights [9, 17, 23, 28, 29]. It was introduced in [27] and subsequently linked with integrable systems and interesting combinatorics [2, 10, 22]. These exactly solvable models are believed to be representative of what should be true more generally.

Organization of the paper. The paper is essentially self-contained. One exception is that in Section 5 we cite variational formulas for the free energy from [24, 25]. Here is an outline of the paper.

Section 2. Introduction of the polymer RWRE in a general context as the positive temperature counterpart of the competition interface of last-passage percolation.

Section 3. Introduction of the log-gamma polymer. The shift-invariant log-gamma polymer is formalized in the definition of a gamma system of weights.

Section 4. Limits of ratios of point-to-point partition functions for the log-gamma polymer.

Section 5. Busemann functions are constructed from limits of ratios of point-to-point partition functions and used to solve a variational formula for the limiting free energy. Duality between tilt and velocity.

Section 6. Limits of ratios of tilted point-to-line partition functions for the log-gamma polymer. Duality between tilt and velocity appears again.

Section 7. Limits of ratios of partition functions yield convergence of polymer measures to the polymer RWRE. The limit RWRE has fluctuations of size $n^{2/3}$ under the averaged measure.

Section 8. A stationary, ergodic distribution for the log-gamma polymer RWRE.

Section 9. Several auxiliary results, including a large deviation bound for the log-gamma polymer and a general ergodic theorem for correctors.

Notation and conventions. $\mathbb{N} = \{1, 2, 3, \dots\}$ and $\mathbb{Z}_+ = \{0, 1, 2, \dots\}$. For $n \in \mathbb{N}$, $[n] = \{1, 2, \dots, n\}$. On \mathbb{R}^2 the ℓ^1 norm is $|x|_1 = |x_1| + |x_2|$, and inequalities are coordinatewise: $(x_1, x_2) \leq (y_1, y_2)$ if $x_r \leq y_r$ for $r \in \{1, 2\}$. Standard basis vectors are $e_1 = (1, 0)$ and $e_2 = (0, 1)$. Our random walks live in \mathbb{Z}_+^2 and admissible paths $x = (x_k)_{k=0}^n$ have steps $z_k = x_k - x_{k-1} \in \{e_1, e_2\}$. Points of \mathbb{Z}_+^2 are written as u, v, x, y but also as (m, n) or (i, j) . Weights indexed by a single point do not have the parentheses: if $x = (i, j)$, then $\eta_x = \eta_{i,j}$. For $u \leq v$ in \mathbb{Z}_+^2 , $\Pi_{u,v}$ is the set of admissible paths from $x_0 = u$ to $x_{|v-u|_1} = v$. Limit velocities of these walks lie in the simplex $\mathcal{U} = \{(u, 1-u) : u \in [0, 1]\}$, whose (relative) interior is denoted by $\text{int } \mathcal{U}$. Shift maps T_v act on suitably indexed configurations $w = (w_x)$ by $(T_v w)_x = w_{v+x}$. \mathbb{E} and \mathbb{P} refer to the random weights (the environment), and otherwise E^μ denotes expectation under probability measure μ . The usual gamma function for $\rho > 0$ is $\Gamma(\rho) = \int_0^\infty x^{\rho-1} e^{-x} dx$, and the Gamma(ρ) distribution on \mathbb{R}_+ is $\Gamma(\rho)^{-1} x^{\rho-1} e^{-x} dx$. $\Psi_0 = \Gamma'/\Gamma$ and $\Psi_1 = \Psi'_0$ are the digamma and trigamma functions.

The reader should be warned that several different partition functions appear in this paper. They are all denoted by Z and sometimes with additional notation such as \check{Z} . It should be clear from the context which Z is meant. Each Z is a sum of weights $W(x_\cdot)$ of paths x_\cdot from a collection of nearest-neighbor lattice paths. Associated to each Z is a polymer probability measure Q on paths: $Q\{x_\cdot\} = Z^{-1}W(x_\cdot)$.

2. THE POLYMER RANDOM WALK IN RANDOM ENVIRONMENT

In this section we introduce a natural random walk in random environment (RWRE) associated to an underlying directed polymer model. This walk appears when we look for the positive temperature counterparts of the notions of geodesics and competition interface that appear in last-passage percolation. Percolation and polymers are discussed in this section in terms of positive weights, without specifying probability distributions.

2.1. Geodesics and competition interface in last-passage percolation. We give a quick definition of last-passage percolation, also known as the zero-temperature polymer. Let $\{\omega_x : x \in \mathbb{Z}_+^2\}$ be a collection of positive weights. For $u \leq v$ in \mathbb{Z}_+^2 let $\Pi_{u,v}$ denote the set of admissible lattice paths $x_\cdot = (x_i)_{0 \leq i \leq n}$ with $n = |v - u|_1$ that satisfy $x_0 = u$, $x_i - x_{i-1} \in \{e_1, e_2\}$, $x_n = v$. The last-passage times are defined by

$$(2.1) \quad G_{u,v} = \max_{x_\cdot \in \Pi_{u,v}} \sum_{i=1}^{|v-u|_1} \omega_{x_i}, \quad u \leq v \text{ in } \mathbb{Z}_+^2.$$

A finite path $(x_i)_{0 \leq i \leq n}$ in $\Pi_{u,v}$ is a *geodesic* between u and v if it is the maximizing path that realizes $G_{u,v}$, namely, $G_{u,v} = \sum_{i=1}^n \omega_{x_i}$. Every subpath of a geodesic is also a geodesic. Let us assume that no two paths of any length have equal sum of weights so that maximizing paths are unique. This would almost surely be the case for example if the weights are i.i.d. with a continuous distribution.

It is convenient to construct the geodesic from u to v backwards, utilizing the iteration

$$(2.2) \quad G_{u,x} = G_{u,x-e_1} \vee G_{u,x-e_2} + \omega_x.$$

Start the construction with $x_n = v$. Suppose the segment $(x_k, x_{k+1}, \dots, x_n)$ of the geodesic has been constructed. If $x_k > u$ coordinatewise, set

$$(2.3) \quad x_{k-1} = \begin{cases} x_k - e_1 & \text{if } G_{u,x_k-e_1} > G_{u,x_k-e_2}, \\ x_k - e_2 & \text{if } G_{u,x_k-e_1} < G_{u,x_k-e_2}. \end{cases}$$

If $x_k \cdot e_r = u \cdot e_r$ for either $r = 1$ or $r = 2$, then define the remaining segment as $(x_0, \dots, x_k) = (u + i e_{3-r})_{0 \leq i \leq k}$.

For a fixed initial point $u \in \mathbb{Z}_+^2$, the *geodesic spanning tree* \mathcal{T}_u of the lattice $u + \mathbb{Z}_+^2$ is the union of all the geodesics from u to v , $v \in u + \mathbb{Z}_+^2$.

The *competition interface* $\varphi = (\varphi_k)_{k \in \mathbb{Z}_+}$ is a lattice path on \mathbb{Z}_+^2 defined as a function of $\{G_{0,v}\}_{v \in \mathbb{Z}_+^2}$. It starts at $\varphi_0 = 0$ and then chooses, at each step, the minimal G -value:

$$(2.4) \quad \varphi_{k+1} = \begin{cases} \varphi_k + e_1 & \text{if } G_{0,\varphi_k+e_1} < G_{0,\varphi_k+e_2}, \\ \varphi_k + e_2 & \text{if } G_{0,\varphi_k+e_1} > G_{0,\varphi_k+e_2}. \end{cases}$$

The relationship between \mathcal{T}_0 and φ is that φ separates the two subtrees $\mathcal{T}_{0,e_1}, \mathcal{T}_{0,e_2}$ of \mathcal{T}_0 rooted at e_1 and e_2 . Since every \mathbb{Z}_+^2 lattice path from 0 has to go through either e_1 or e_2 , $\mathcal{T}_0 = \{0\} \cup \mathcal{T}_{0,e_1} \cup \mathcal{T}_{0,e_2}$ as a disjoint union. For each $n \in \mathbb{Z}_+$, φ_n is the unique point such that $|\varphi_n|_1 = n$ and for $r \in \{1, 2\}$, $\{\varphi_n + ke_r : k \in \mathbb{N}\} \subseteq \mathcal{T}_{0,e_r}$. Note that we cannot say which tree contains φ_n , unless we know that $\varphi_n - \varphi_{n-1} = e_r$ in which case $\varphi_n \in \mathcal{T}_{0,e_r}$. If we shift φ by $(1/2, 1/2)$ then it threads exactly between the two trees (Figure 1).

The term competition interface comes from the interpretation that \mathcal{T}_{0,e_1} and \mathcal{T}_{0,e_2} are two competing clusters or infections on the lattice [13, 14]. The model can be defined dynamically. The clusters at time $t \in \mathbb{R}_+$ are $\mathcal{T}_{0,e_r}(t) = \{v \in \mathcal{T}_{0,e_r} : G_{0,v} \leq t\}$.

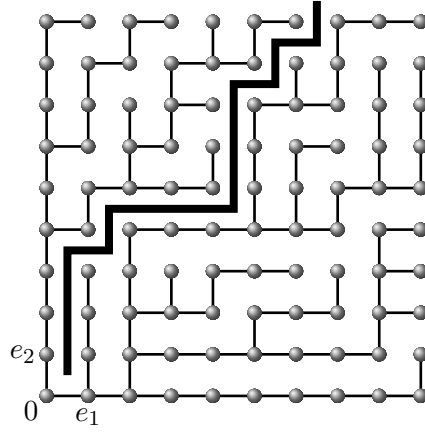


FIGURE 1. The competition interface shifted by $(1/2, 1/2)$ (solid line) separating the subtrees of \mathcal{T}_0 rooted at e_1 and e_2 .

2.2. Geodesics and competition interface in positive temperature. We switch to a positive temperature polymer. Let $\{V_x\}_{x \in \mathbb{Z}_+^2}$ be positive weights. Define point-to-point polymer partition functions for $u \leq v$ in \mathbb{Z}_+^2 by

$$(2.5) \quad Z_{u,v} = \sum_{x \in \Pi_{u,v}} \prod_{i=1}^{|v-u|_1} V_{x_i}^{-1}$$

and the polymer measure on the set of paths $\Pi_{u,v}$ by

$$(2.6) \quad Q_{u,v}\{x\} = \frac{1}{Z_{u,v}} \prod_{i=1}^{|v-u|_1} V_{x_i}^{-1}, \quad x \in \Pi_{u,v}.$$

Our convention is to use reciprocals V_x^{-1} of the weights in the definitions. The reason is that this way the weights in the log-gamma polymer are gamma distributed and features of the beta-gamma algebra arise naturally.

Remark 2.1. A conventional way of defining polymer partition functions is

$$(2.7) \quad Z_{u,v}^\beta = \sum_{x \in \Pi_{u,v}} e^{\beta \sum_{i=1}^{|v-u|_1} \omega_{x_i}}$$

with an inverse temperature parameter $0 < \beta < \infty$. In the zero-temperature limit $\beta^{-1} \log Z_{u,v}^\beta \rightarrow G_{u,v}$ as $\beta \rightarrow \infty$, and the polymer measure $Q_{u,v}^\beta$ concentrates on the geodesic(s) from u to v . This is the sense in which last-passage percolation is the zero-temperature polymer. See Remark 3.2 below for this point for the log-gamma polymer.

We implement noisy versions of rules (2.3) and (2.4) to define positive temperature counterparts of geodesics and the competition interface.

Fix a base point $u \in \mathbb{Z}_+^2$ and define a backward Markov transition kernel $\overleftarrow{\pi}^u$ on the lattice $u + \mathbb{Z}_+^2$ by $\overleftarrow{\pi}^u(u, u) = 1$, and

$$(2.8) \quad \overleftarrow{\pi}^u(x, x - e_r) = \frac{V_x^{-1} Z_{u, x - e_r}}{Z_{u, x}} = \frac{Z_{u, x - e_r}}{Z_{u, x - e_1} + Z_{u, x - e_2}}, \quad \text{for } r \in \{1, 2\},$$

if both x and $x - e_r$ lie in $u + \mathbb{Z}_+^2$. The middle formula above gives the correct values on the boundaries of $u + \mathbb{Z}_+^2$ where there is only one admissible backward step: $\overleftarrow{\pi}^u(u + i e_r, u + (i - 1) e_r) = 1$ for $i \geq 1$ and $r \in \{1, 2\}$.

For a path $x_\cdot \in \Pi_{u,v}$ comparison of (2.6) and (2.8) shows

$$Q_{u,v}\{x_\cdot\} = \prod_{i=1}^{|v-u|_1} \overleftarrow{\pi}^u(x_i, x_{i-1}).$$

So the quenched polymer distribution $Q_{u,v}$ is the distribution of the backward Markov chain with initial state v , transition $\overleftarrow{\pi}^u$, and absorption at u . The distributions $Q_{u,v}$ are the noisy counterparts of geodesics. The nesting property of geodesics manifests itself through conditioning. Let $u < z < w < v$ in \mathbb{Z}_+^2 . Let $A_{z,w}$ be the set of paths in $\Pi_{u,v}$ that go through the points z and w . Given $y_\cdot \in \Pi_{z,w}$, let B_{y_\cdot} be the set of paths in $\Pi_{u,v}$ that traverse the path y_\cdot (that is, contain y_\cdot as a subpath). Then

$$(2.9) \quad Q_{u,v}(B_{y_\cdot} | A_{z,w}) = Q_{z,w}\{y_\cdot\}.$$

Define the random geodesic spanning tree \mathcal{T}_u rooted at u by choosing, for each $x \in (u + \mathbb{Z}_+^2) \setminus \{u\}$, a parent

$$(2.10) \quad \gamma(x) = \begin{cases} x - e_1 & \text{with probability } \overleftarrow{\pi}^u(x, x - e_1) \\ x - e_2 & \text{with probability } \overleftarrow{\pi}^u(x, x - e_2). \end{cases}$$

Now that we have the positive temperature counterparts of geodesics, we can find the positive temperature counterpart of the competition interface, by reference to the tree \mathcal{T}_0 rooted at 0. Let \mathcal{T}_{0,e_r} be the subtree rooted at e_r , so that $\mathcal{T}_0 = \{0\} \cup \mathcal{T}_{0,e_1} \cup \mathcal{T}_{0,e_2}$ as a disjoint union. The lemma below shows that there is a well-defined path X_\cdot that separates the trees \mathcal{T}_{0,e_1} and \mathcal{T}_{0,e_2} , and evolves as a Markov chain in the environment defined by the partition functions. In other words, this random walk in a random environment (RWRE) is the positive temperature analogue of the competition interface. The picture for X_\cdot is the same as for φ in Figure 1.

LEMMA 2.2. (a) *Given the choices made in (2.10), there is a unique lattice path $(X_n)_{n \in \mathbb{Z}_+}$ with these properties: $X_0 = 0$, $X_n - X_{n-1} \in \{e_1, e_2\}$, and for each n and $r \in \{1, 2\}$, $\{X_n + ke_r : k \in \mathbb{N}\} \subseteq \mathcal{T}_{0, e_r}$.*

(b) *X_n is a Markov chain with transition matrix*

$$(2.11) \quad \pi_{x, x+e_r} = \frac{Z_{0, x+e_r}^{-1}}{Z_{0, x+e_1}^{-1} + Z_{0, x+e_2}^{-1}}, \quad x \in \mathbb{Z}_+^2, \quad r \in \{1, 2\}.$$

Proof. (a) To prove the existence of the path, start with $X_0 = 0$, and iterate the following move: if $\gamma(X_n + e_1 + e_2) = X_n + e_r$ set $X_{n+1} = X_n + e_{3-r}$.

(b) Given the path $(X_k)_{k=0}^n$ with $X_n = x$, we choose $X_{n+1} = x + e_1$ if $\gamma(x + e_1 + e_2) = x + e_2$ which happens with probability

$$\frac{Z_{0, x+e_2}}{Z_{0, x+e_1} + Z_{0, x+e_2}} = \frac{Z_{0, x+e_1}^{-1}}{Z_{0, x+e_1}^{-1} + Z_{0, x+e_2}^{-1}}.$$

Similarly for $X_{n+1} = x + e_2$ with the complementary probability. \square

A genuine RWRE transition probability should satisfy $\pi_{x,y}(\omega) = \pi_{0,y-x}(T_x \omega)$ for shift mappings $(T_x)_{x \in \mathbb{Z}_+^2}$ acting on the environments ω . We augment the space of weights to achieve this. We need to be precise about the sets of sites on which various classes of weights are defined.

Definition 2.3. A collection of positive real weights $(\xi, \eta, \zeta, \check{\xi}) = \{\xi_x, \eta_{x-e_2}, \zeta_{x-e_1}, \check{\xi}_{x-e_1-e_2} : x \in \mathbb{N}^2\}$ satisfies *north-east (NE) induction* if these equations hold for each $x \in \mathbb{N}^2$:

$$(2.12) \quad \eta_x = \xi_x \frac{\eta_{x-e_2}}{\eta_{x-e_2} + \zeta_{x-e_1}}, \quad \zeta_x = \xi_x \frac{\zeta_{x-e_1}}{\eta_{x-e_2} + \zeta_{x-e_1}},$$

$$(2.13) \quad \text{and} \quad \check{\xi}_{x-e_1-e_2} = \eta_{x-e_2} + \zeta_{x-e_1}.$$

North-east induction simply keeps track of ratios of partition functions. Take as given the subcollection of weights $\{\xi_{i,j}, \eta_{i,0}, \zeta_{0,j} : i, j \in \mathbb{N}\}$ on \mathbb{Z}_+^2 and construct the polymer partition functions

$$(2.14) \quad Z_{0,v} = \sum_{x \in \Pi_{0,v}} \prod_{i=1}^{|v|_1} V_{x_i}^{-1} \quad \text{with} \quad V_{i,j} = \begin{cases} \xi_{i,j}, & (i,j) \in \mathbb{N}^2 \\ \eta_{i,0}, & i \in \mathbb{N}, j = 0 \\ \zeta_{0,j}, & i = 0, j \in \mathbb{N}. \end{cases}$$

Then define

$$(2.15) \quad \eta_x = \frac{Z_{0, x-e_1}}{Z_{0,x}} \quad \text{and} \quad \zeta_x = \frac{Z_{0, x-e_2}}{Z_{0,x}} \quad \text{for } x \in \mathbb{N}^2.$$

Now the subsystem (ξ, η, ζ) satisfies (2.12), as can be verified by induction. To get the full system $(\xi, \eta, \zeta, \check{\xi})$ just define $\check{\xi}_x = \eta_{x+e_1} + \zeta_{x+e_2}$ for $x \in \mathbb{Z}_+^2$.

Note that (2.15) is valid also on the boundaries, by the definition (2.14) of Z_{0, ke_r} for $k \in \mathbb{N}$. The reason for the distinct notation $\{\eta_{i,0}, \zeta_{0,j}\}$ for boundary weights in (2.14) is that these are also ratios of partition functions, just as η_x and ζ_x in (2.15). We shall find

that in the interesting log-gamma models the boundary weights $\{\eta_{i,0}, \zeta_{0,j}\}$ are different from the bulk weights $\{\xi_{i,j}\}$. The role of the $\check{\xi}$ weights is not clear yet, but they will become central in the log-gamma context.

Define the space of environments

$$(2.16) \quad \Omega_{\text{NE}} = \{\omega = (\xi, \eta, \zeta, \check{\xi}) \in \mathbb{R}_+^{\mathbb{N}^2 + (\mathbb{N} \times \mathbb{Z}_+) + (\mathbb{Z}_+ \times \mathbb{N}) + \mathbb{Z}_+^2} : (\xi, \eta, \zeta, \check{\xi}) \text{ satisfies NE induction}\}.$$

Translations act naturally: for $z \in \mathbb{Z}_+^2$, $T_z \omega = (\xi_{z+\mathbb{N}^2}, \eta_{z+\mathbb{N} \times \mathbb{Z}_+}, \zeta_{z+\mathbb{Z}_+ \times \mathbb{N}}, \check{\xi}_{z+\mathbb{Z}_+^2})$ where we introduced notation $\xi_{z+\mathbb{N}^2} = \{\xi_{z+x}\}_{x \in \mathbb{N}^2}$, and similarly for the other configurations.

Definition 2.4. The *polymer random walk in random environment* is a RWRE with environment space Ω_{NE} and transition probability

$$(2.17) \quad \pi_{x, x+e_1}(\omega) = \frac{\eta_{x+e_1}}{\eta_{x+e_1} + \zeta_{x+e_2}} \quad \text{and} \quad \pi_{x, x+e_2}(\omega) = \frac{\zeta_{x+e_2}}{\eta_{x+e_1} + \zeta_{x+e_2}}.$$

This definition is the same as (2.11) with partition functions (2.14). The quenched path probabilities P^ω of this RWRE started at 0 are defined by

$$(2.18) \quad P^\omega(X_0 = 0, X_1 = x_1, \dots, X_n = x_n) = \prod_{k=1}^n \pi_{x_{k-1}, x_k}(\omega) \quad \text{with } x_0 = 0.$$

Distributions of X_n are again related to polymer distributions and partition functions. Define

$$(2.19) \quad \check{Z}_{0,v} = \sum_{x \in \Pi_{0,v}} \prod_{i=0}^{|v|_1-1} \check{\xi}_{x_i}^{-1}, \quad v \in \mathbb{Z}_+^2.$$

In contrast with (2.14), this time the weight at the origin is included but the weight at v excluded. From (2.13) and (2.18) we derive two formulas. First, distribution of X_n is a ratio of partition functions:

$$(2.20) \quad P^\omega(X_n = x) = \frac{\check{Z}_{0,x}}{Z_{0,x}} \quad \text{for } x \in \mathbb{Z}_+^2 \text{ such that } |x|_1 = n.$$

Then, if the walk is conditioned to go through a point, the distribution of the path segment is the polymer probability in $\check{\xi}$ weights: for $x_\cdot = (x_k)_{k=0}^n \in \Pi_{0,x_n}$,

$$(2.21) \quad P^\omega(X_0 = 0, X_1 = x_1, \dots, X_n = x_n | X_n = x_n) = \frac{1}{\check{Z}_{0,x_n}} \prod_{i=0}^{n-1} \check{\xi}_{x_i}^{-1} = \check{Q}_{0,x_n}\{x_\cdot\}.$$

3. THE LOG-GAMMA POLYMER

This section gives a quick definition of the log-gamma polymer and its Burke property. Let $0 < \lambda < \rho < \infty$.

Definition 3.1. A collection $(\xi, \eta, \zeta, \check{\xi}) = \{\xi_x, \eta_{x-e_2}, \zeta_{x-e_1}, \check{\xi}_{x-e_1-e_2} : x \in \mathbb{N}^2\}$ of positive real random variables is a *gamma system* of weights with parameters (λ, ρ) if the following three properties hold.

(a) NE induction (Definition 2.3) holds: for each $x \in \mathbb{N}^2$, almost surely,

$$(3.1) \quad \eta_x = \xi_x \frac{\eta_{x-e_2}}{\eta_{x-e_2} + \zeta_{x-e_1}}, \quad \zeta_x = \xi_x \frac{\zeta_{x-e_1}}{\eta_{x-e_2} + \zeta_{x-e_1}},$$

and $\check{\xi}_{x-e_1-e_2} = \eta_{x-e_2} + \zeta_{x-e_1}.$

(b) The marginal distributions of the variables are

$$(3.2) \quad \eta_x \sim \text{Gamma}(\lambda), \zeta_x \sim \text{Gamma}(\rho - \lambda), \text{ and } \xi_x, \check{\xi}_x \sim \text{Gamma}(\rho).$$

(c) The variables $\{\xi_{i,j}, \eta_{i,0}, \zeta_{0,j} : i, j \in \mathbb{N}\}$ are mutually independent.

A triple (ξ, η, ζ) is a gamma system with parameters (λ, ρ) if conditions (a)–(c) are satisfied without the conditions on $\check{\xi}$.

Note that the ξ -weights are defined only in the bulk \mathbb{N}^2 , while the $\check{\xi}$ -weights are defined also on the boundaries and at the origin. Variables η_x and ζ_x can be thought of as weights on the edges, while ξ_x and $\check{\xi}_x$ are weights on the vertices. Edge weights will also be denoted by

$$(3.3) \quad \tau_{x-e_1,x} = \eta_x \quad \text{and} \quad \tau_{x-e_2,x} = \zeta_x.$$

A natural way to think about equations (3.1) for a fixed x is as a mapping of triples (Figure 2):

$$(3.4) \quad (\xi_x, \eta_{x-e_2}, \zeta_{x-e_1}) \mapsto (\eta_x, \zeta_x, \check{\xi}_{x-e_1-e_2}).$$

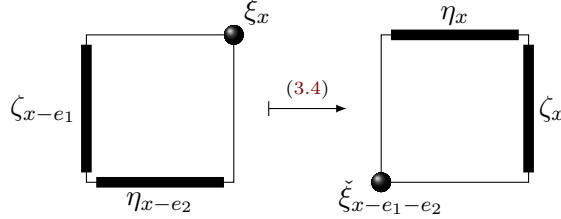


FIGURE 2. Mapping (3.4) that involves variables on a single lattice square. The picture illustrates how southwest corners are flipped into northeast corners in an inductive proof of the Burke property.

This mapping preserves the independent gamma distributions: if $(\xi_x, \eta_{x-e_2}, \zeta_{x-e_1})$ are independent with marginals (3.2), then the same is true for $(\eta_x, \zeta_x, \check{\xi}_{x-e_1-e_2})$, as can be checked for example via Laplace transforms. Consequently a gamma system $(\xi, \eta, \zeta, \check{\xi})$ can be constructed by repeated application of equations (3.1) to independent gamma variables given in (c).

An equivalent way to define the gamma system is to first construct the following polymer partition functions from the weights given in (c): for $0 \leq u < v$ in \mathbb{Z}_+^2 ,

$$(3.5) \quad Z_{u,v} = \sum_{x \in \Pi_{u,v}} \prod_{i=1}^{|v-u|_1} V_{x_i}^{-1} \quad \text{with} \quad V_{i,j} = \begin{cases} \xi_{i,j}, & (i,j) \in \mathbb{N}^2 \\ \eta_{i,0}, & i \in \mathbb{N}, j = 0 \\ \zeta_{0,j}, & i = 0, j \in \mathbb{N}. \end{cases}$$

Then define

$$(3.6) \quad \tau_{x-e_r, x} = \frac{Z_{0, x-e_r}}{Z_{0, x}} \quad \text{for } x \in \mathbb{Z}_+^2 \text{ and } r \in \{1, 2\} \text{ such that } x - e_r \in \mathbb{Z}_+^2.$$

The weights η_x and ζ_x are then defined via (3.3). Now we have a gamma system (ξ, η, ζ) , which can be augmented to a gamma system $(\xi, \eta, \zeta, \check{\xi})$ since $\check{\xi}$ is a function of (η, ζ) .

Mapping (3.4) furnishes the induction step in the proof of the *Burke property* of the log-gamma polymer [27, Thm. 3.3]: for any down-right path on \mathbb{Z}_+^2 , the τ -variables on the path, the ξ variables strictly to the northeast of the path, and the $\check{\xi}$ variables strictly to the southwest of the path are all mutually independent with marginal distributions (3.2). The induction proof begins with the path that consists of the e_1 - and e_2 -axes. Southwest corners of the path can be flipped into northeast corners by an application of (3.4), as illustrated in Figure 2.

As an application of the Burke property consider the down-right path consisting of the north and east boundaries of the rectangle $\{0, \dots, m\} \times \{0, \dots, n\}$. Then the Burke property gives us this statement:

$$(3.7) \quad \begin{aligned} &\text{variables } \{\eta_{i,n}, \zeta_{m,j}, \check{\xi}_{i-1,j-1} : 1 \leq i \leq m, 1 \leq j \leq n\} \\ &\text{are mutually independent with marginals (3.2).} \end{aligned}$$

Remark 3.2. Let us revisit the zero temperature limit (Remark 2.1). The log-gamma polymer does not have an explicit β parameter but ρ represents temperature. Replace ρ by $\varepsilon\rho$ in the definitions above, so that $\xi_x \sim \text{Gamma}(\varepsilon\rho)$. Then as $\varepsilon \searrow 0$, $-\varepsilon \log \xi_x \Rightarrow \omega_x$, a rate ρ exponential weight. For $u < v$ in \mathbb{N} ,

$$\begin{aligned} \varepsilon \log Z_{u,v} &= \varepsilon \log \sum_{x \in \Pi_{u,v}} \exp \left\{ -\varepsilon^{-1} \sum_{i=1}^{|v-u|_1} \varepsilon \log \xi_{x_i} \right\} \\ &\Rightarrow \max_{x \in \Pi_{u,v}} \sum_{i=1}^{|v-u|_1} \omega_x = G_{u,v} \quad \text{as } \varepsilon \searrow 0. \end{aligned}$$

In other words, we have convergence in distribution to last-passage percolation with exponential weights.

An important function of the polymer path is the exit point or exit time t_{exit} of the path from the boundary: $t_{\text{exit}} = t_{e_1} \vee t_{e_2}$,

$$(3.8) \quad t_{e_1} = \max\{k \geq 0 : x_i = (i, 0) \text{ for } 0 \leq i \leq k\}$$

and

$$(3.9) \quad t_{e_2} = \max\{\ell \geq 0 : x_j = (0, j) \text{ for } 0 \leq j \leq \ell\}.$$

Note that for each path $t_{e_1} \wedge t_{e_2} = 0$. Partition functions (3.5) based at 0 can be equivalently written as

$$(3.10) \quad Z_{0,v} = \sum_{x \in \Pi_{0,v}} \left(\prod_{i=1}^{t_{\text{exit}}} \tau_{x_{i-1}, x_i}^{-1} \right) \left(\prod_{j=t_{\text{exit}}+1}^{|v|_1} \xi_{x_j}^{-1} \right), \quad v \in \mathbb{Z}_+^2.$$

In a (λ, ρ) gamma system we have the means $\mathbb{E}(\log \eta_{i,0}) = \Psi_0(\lambda)$ and

$$(3.11) \quad \mathbb{E}(\log Z_{0,(m,n)}) = -m\Psi_0(\lambda) - n\Psi_0(\rho - \lambda).$$

The second one comes from

$$(3.12) \quad \log Z_{0,(m,n)} = -\sum_{i=1}^m \log \eta_{i,0} - \sum_{j=1}^n \log \zeta_{m,j},$$

a sum of two correlated sums of i.i.d. random variables. Above $\Psi_0 = \Gamma'/\Gamma$ is the digamma function. It is strictly increasing on $(0, \infty)$, with $\Psi_0(0+) = -\infty$ and $\Psi_0(\infty) = \infty$. Its derivative is the trigamma function $\Psi_1 = \Psi_0'$ that is convex, strictly decreasing, with $\Psi_1(0+) = \infty$ and $\Psi_1(\infty) = 0$.

The asymptotic directions (or velocities) of admissible paths in \mathbb{Z}_+^2 lie in the simplex $\mathcal{U} = \{\mathbf{u} = (u, 1-u) : u \in [0, 1]\}$. Fundamental to the behavior of the log-gamma polymer is a 1-1 correspondence between velocities $\mathbf{u} \in \mathcal{U}$ and parameters $\lambda \in [0, \rho]$, for a fixed ρ . The *characteristic direction* for (λ, ρ) is

$$(3.13) \quad \mathbf{u}_{\lambda,\rho} = \left(\frac{\Psi_1(\rho - \lambda)}{\Psi_1(\lambda) + \Psi_1(\rho - \lambda)}, \frac{\Psi_1(\lambda)}{\Psi_1(\lambda) + \Psi_1(\rho - \lambda)} \right) \in \mathcal{U}.$$

Conversely, for $\mathbf{u} = (u, 1-u)$, the unique parameter $\theta(u) = \theta(\mathbf{u}) \in [0, \rho]$ for which \mathbf{u} is the characteristic direction is defined by $\theta(0) = 0$, $\theta(1) = \rho$, and

$$(3.14) \quad -u\Psi_1(\theta(u)) + (1-u)\Psi_1(\rho - \theta(u)) = 0 \quad \text{for } u \in (0, 1).$$

Function $\theta(u)$ is a strictly increasing bijective mapping between $u \in [0, 1]$ and $\theta \in [0, \rho]$.

The function $\theta(\mathbf{u})$ will appear throughout the paper. Let us point out that if $(m, n) = c\mathbf{u}$ then the right-hand side of (3.11) is minimized by $\lambda = \theta(\mathbf{u})$. As we shall see, this identifies the limiting free energy for the log-gamma polymer with i.i.d. Gamma(ρ) weights. Notationally, λ, α, ν denote generic parameters in $[0, \rho]$, while θ is reserved for the function defined above.

4. LIMITS OF RATIOS OF POINT-TO-POINT PARTITION FUNCTIONS

Fix $0 < \rho < \infty$. Let i.i.d. Gamma(ρ) weights $w = \{w_x : x \in \mathbb{Z}_+^2\}$ be given. Define partition functions

$$(4.1) \quad Z_{u,v} = \sum_{x \in \Pi_{u,v}} \prod_{i=0}^{|v-u|_1-1} w_{x_i}^{-1}, \quad 0 \leq u \leq v \text{ in } \mathbb{Z}_+^2.$$

Note that the weight at u is included and v excluded, in contrast with definitions (2.5) and (3.5). This is for convenience, to have clean limit statements below.

Suppose a lattice point $(m, n) \in \mathbb{N}^2$ tends to infinity in the first quadrant so that it has an asymptotic direction in the interior of the quadrant. Let $\lambda \in (0, \rho)$ be the unique value such that this assumption holds:

$$(4.2) \quad m \wedge n \rightarrow \infty \quad \text{and} \quad \frac{m}{n} \rightarrow \frac{\Psi_1(\rho - \lambda)}{\Psi_1(\lambda)}.$$

When (4.2) holds we say that $(m, n) \rightarrow \infty$ in the characteristic direction of (λ, ρ) .

The central theorem of this paper constructs gamma systems out of i.i.d. weights by taking limits of ratios of point-to-point partition functions.

THEOREM 4.1. *On the probability space of the i.i.d. Gamma(ρ) weights $w = \{w_x : x \in \mathbb{Z}_+^2\}$ there exist random variables $\{\xi_x^\lambda, \eta_{x-e_2}^\lambda, \zeta_{x-e_1}^\lambda : \lambda \in (0, \rho), x \in \mathbb{N}^2\}$ with the following properties.*

(i) *For each $\lambda \in (0, \rho)$, $(\xi^\lambda, \eta^\lambda, \zeta^\lambda, w)$ is a gamma system with parameters (λ, ρ) . Furthermore, if on the same probability space there are additional random variables $(\tilde{\xi}, \tilde{\eta}, \tilde{\zeta}) = \{\tilde{\xi}_x, \tilde{\eta}_{x-e_2}, \tilde{\zeta}_{x-e_1} : x \in \mathbb{N}^2\}$ such that $(\tilde{\xi}, \tilde{\eta}, \tilde{\zeta}, w)$ is a gamma system with parameters (ν, ρ) , then $(\tilde{\xi}, \tilde{\eta}, \tilde{\zeta}) = (\xi^\nu, \eta^\nu, \zeta^\nu)$ a.s.*

(ii) *If a sequence $(m, n) \rightarrow \infty$ in the characteristic direction of (λ, ρ) , as defined in (4.2), then these almost sure limits hold:*

$$(4.3) \quad \eta_x^\lambda = \lim_{(m,n) \rightarrow \infty} \frac{Z_{x,(m,n)}}{Z_{x-e_1,(m,n)}} \quad \text{and} \quad \zeta_y^\lambda = \lim_{(m,n) \rightarrow \infty} \frac{Z_{y,(m,n)}}{Z_{y-e_2,(m,n)}}$$

for all $x \in \mathbb{N} \times \mathbb{Z}_+$ and $y \in \mathbb{Z}_+ \times \mathbb{N}$.

(iii) *The weights are continuous in λ , and the edge weights are monotone in λ : for each x for which the weights are defined, almost surely,*

$$(4.4) \quad \eta_x^{\lambda_1} \leq \eta_x^{\lambda_2} \quad \text{and} \quad \zeta_x^{\lambda_1} \geq \zeta_x^{\lambda_2} \quad \text{for } \lambda_1 \leq \lambda_2$$

and

$$(4.5) \quad \eta_x^\lambda \rightarrow \eta_x^\nu, \quad \zeta_x^\lambda \rightarrow \zeta_x^\nu, \quad \xi_x^\lambda \rightarrow \xi_x^\nu \quad \text{as } \lambda \rightarrow \nu.$$

The rest of this section proves Theorem 4.1. The reader not interested in the (rather technical) proof can proceed to the next section where these limits are applied to solve a variational problem for the limiting free energy.

The proof relies on the following lemma for gamma systems. Let $(\xi, \eta, \zeta, \check{\xi})$ be an (α, ρ) -system according to Definition 3.1. Using the $\check{\xi}$ weights, define partition functions

$$(4.6) \quad \check{Z}_{u,v} = \sum_{x \in \Pi_{u,v}} \prod_{i=0}^{|v-u|-1} (\check{\xi}_{x_i})^{-1}, \quad 0 \leq u \leq v \text{ in } \mathbb{Z}_+^2,$$

and for $x \in \mathbb{N} \times \mathbb{Z}_+$ and $y \in \mathbb{Z}_+ \times \mathbb{N}$ edge ratio weights

$$(4.7) \quad \check{I}_{x,(m,n)} = \frac{\check{Z}_{x,(m,n)}}{\check{Z}_{x-e_1,(m,n)}} \quad \text{and} \quad \check{J}_{y,(m,n)} = \frac{\check{Z}_{y,(m,n)}}{\check{Z}_{y-e_2,(m,n)}}.$$

LEMMA 4.2. *Let $0 < \lambda < \alpha < \tilde{\lambda} < \rho$. Consider two sequences $(m_i, n_i) \rightarrow \infty$ and $(\tilde{m}_j, \tilde{n}_j) \rightarrow \infty$ in \mathbb{N}^2 such that*

$$(4.8) \quad \frac{m_i}{n_i} \rightarrow \frac{\Psi_1(\rho - \lambda)}{\Psi_1(\lambda)} \quad \text{and} \quad \frac{\tilde{m}_j}{\tilde{n}_j} \rightarrow \frac{\Psi_1(\rho - \tilde{\lambda})}{\Psi_1(\tilde{\lambda})}.$$

Then for $x \in \mathbb{N} \times \mathbb{Z}_+$ and $y \in \mathbb{Z}_+ \times \mathbb{N}$

$$(4.9) \quad \overline{\lim}_{i \rightarrow \infty} \check{I}_{x,(m_i,n_i)} \leq \eta_x \leq \underline{\lim}_{j \rightarrow \infty} \check{I}_{x,(\tilde{m}_j,\tilde{n}_j)} \quad \text{a.s.}$$

and

$$(4.10) \quad \overline{\lim}_{j \rightarrow \infty} \check{J}_{y,(\tilde{m}_j, \tilde{n}_j)} \leq \zeta_y \leq \underline{\lim}_{i \rightarrow \infty} \check{J}_{y,(m_i, n_i)} \quad a.s.$$

Proof of Lemma 4.2. For notational simplicity we drop the i, j indices from (m, n) and (\tilde{m}, \tilde{n}) . We relate ratios (4.7) to ratios of partition functions with boundaries. Let $Z_{(k, \ell), (m, n)}^{\text{NE}}$ denote a partition function that uses η and ζ weights on the north and east boundaries of the rectangle $\{k, \dots, m\} \times \{\ell, \dots, n\}$ and ξ weights in the bulk:

$$Z_{(k, n), (m, n)}^{\text{NE}} = \prod_{s=k+1}^m \frac{1}{\eta_{s, n}}, \quad Z_{(m, \ell), (m, n)}^{\text{NE}} = \prod_{t=\ell+1}^n \frac{1}{\zeta_{m, t}}$$

and for $0 \leq k < m$ and $0 \leq \ell < n$

$$(4.11) \quad \begin{aligned} Z_{(k, \ell), (m, n)}^{\text{NE}} &= \sum_{i=k}^{m-1} \check{Z}_{(k, \ell), (i, n-1)} \frac{1}{\check{\xi}_{i, n-1}} \prod_{s=i+1}^m \frac{1}{\eta_{s, n}} \\ &\quad + \sum_{j=\ell}^{n-1} \check{Z}_{(k, \ell), (m-1, j)} \frac{1}{\check{\xi}_{m-1, j}} \prod_{t=j+1}^n \frac{1}{\zeta_{m, t}}. \end{aligned}$$

In the last formula $Z_{(k, \ell), (m, n)}^{\text{NE}}$ is decomposed according to the entry points (i, n) and (m, j) of the paths on the north and east boundaries. If the entry is at (i, n) , the first boundary variable encountered is $\eta_{i+1, n}$ associated to the edge $\{(i, n), (i+1, n)\}$. The last bulk weight $\check{\xi}_{i, n-1}$ has to be inserted explicitly into the formula because $\check{Z}_{(k, \ell), (i, n)}$ does not include the weight at (i, n) , by its definition (4.6).

The corresponding ratio weights on edges are

$$(4.12) \quad I_{(k, \ell), (m, n)} = \frac{Z_{(k, \ell), (m, n)}^{\text{NE}}}{Z_{(k-1, \ell), (m, n)}^{\text{NE}}} \quad \text{and} \quad J_{(k, \ell), (m, n)} = \frac{Z_{(k, \ell), (m, n)}^{\text{NE}}}{Z_{(k, \ell-1), (m, n)}^{\text{NE}}}.$$

Due to the reversibility of the shift-invariant setting, these ratio weights are the same as the original ratio weights, and thereby do not depend on (m, n) . This is the content of the next lemma.

LEMMA 4.3. *For $0 \leq k \leq m$ and $0 \leq \ell \leq n$ such that the weights below are defined,*

$$(4.13) \quad \eta_{k, \ell} = I_{(k, \ell), (m, n)} \quad \text{and} \quad \zeta_{k, \ell} = J_{(k, \ell), (m, n)}.$$

Proof of Lemma 4.3. When $\ell = n$ in the η -identity or $k = m$ in the ζ -identity the claims follow from the definitions. Here is the induction step for $\eta_{k, \ell}$, assuming the identities have been verified for the edges $\{(k-1, \ell+1), (k, \ell+1)\}$ and $\{(k, \ell), (k, \ell+1)\}$, closest to the

north and east of the edge $\{(k-1, \ell), (k, \ell)\}$:

$$\begin{aligned}
I_{(k, \ell), (m, n)} &= \frac{\check{\xi}_{k-1, \ell} Z_{(k, \ell), (m, n)}^{\text{NE}}}{Z_{(k, \ell), (m, n)}^{\text{NE}} + Z_{(k-1, \ell+1), (m, n)}^{\text{NE}}} \\
&= \check{\xi}_{k-1, \ell} \left(1 + \frac{Z_{(k-1, \ell+1), (m, n)}^{\text{NE}}}{Z_{(k, \ell+1), (m, n)}^{\text{NE}}} \cdot \frac{Z_{(k, \ell+1), (m, n)}^{\text{NE}}}{Z_{(k, \ell), (m, n)}^{\text{NE}}} \right)^{-1} \\
&= \check{\xi}_{k-1, \ell} \left(1 + \frac{\zeta_{k, \ell+1}}{\eta_{k, \ell+1}} \right)^{-1} = (\eta_{k, \ell} + \zeta_{k-1, \ell+1}) \left(1 + \frac{\zeta_{k-1, \ell+1}}{\eta_{k, \ell}} \right)^{-1} = \eta_{k, \ell}.
\end{aligned}$$

The third equality is the induction step. The fourth equality uses (3.1) twice. \square

We need one more variant of ratio weights, namely the types where the last step of the path is restricted to either e_1 or e_2 . Relative to any fixed rectangle $\{k, \dots, m\} \times \{\ell, \dots, n\}$, define the distances of the entrance points of the polymer path $x. \in \Pi_{(k, \ell), (m, n)}$ on the north and east boundaries to the corner (m, n) :

$$(4.14) \quad t_{e_1}^* = \max\{r \geq 0 : x_{m-k+n-\ell-i} = (m-i, n) \text{ for } 0 \leq i \leq r\}$$

and

$$(4.15) \quad t_{e_2}^* = \max\{r \geq 0 : x_{m-k+n-\ell-j} = (m, n-j) \text{ for } 0 \leq j \leq r\}.$$

For a subset A of paths, write $Z(A)$ for the partition function of paths restricted to A (in other words, for the unnormalized polymer measure). Then define, for $r \in \{1, 2\}$,

$$\begin{aligned}
(4.16) \quad I_{(k, \ell), (m, n)}^{e_r} &= \frac{Z_{(k, \ell), (m, n)}^{\text{NE}}(t_{e_r}^* > 0)}{Z_{(k-1, \ell), (m, n)}^{\text{NE}}(t_{e_r}^* > 0)} \\
\text{and} \quad J_{(k, \ell), (m, n)}^{e_r} &= \frac{Z_{(k, \ell), (m, n)}^{\text{NE}}(t_{e_r}^* > 0)}{Z_{(k, \ell-1), (m, n)}^{\text{NE}}(t_{e_r}^* > 0)}.
\end{aligned}$$

We are ready to prove Lemma 4.2. We go through the proof of (4.9), the case for \check{J} being the same. Applying Lemma 9.1 from the appendix (to a reversed rectangle) gives

$$(4.17) \quad \eta_{k, \ell} + (I_{(k, \ell), (m+1, n+1)}^{e_1} - \eta_{k, \ell}) \leq \check{I}_{(k, \ell), (m, n)} \leq \eta_{k, \ell} + (I_{(k, \ell), (m+1, n+1)}^{e_2} - \eta_{k, \ell}).$$

Taking (4.13) into consideration, the task is

$$\begin{aligned}
(4.18) \quad &\overline{\lim}_{(m, n) \rightarrow \infty} \{I_{(k, \ell), (m+1, n+1)}^{e_2} - I_{(k, \ell), (m+1, n+1)}\} \\
&\leq 0 \leq \underline{\lim}_{(\tilde{m}, \tilde{n}) \rightarrow \infty} \{I_{(k, \ell), (\tilde{m}+1, \tilde{n}+1)}^{e_1} - I_{(k, \ell), (\tilde{m}+1, \tilde{n}+1)}\}.
\end{aligned}$$

We do the first limit for e_2 . The second is similar. Introduce the parameter

$$N = \frac{m+n}{\Psi_1(\rho-\lambda) + \Psi_1(\lambda)} \rightarrow \infty \quad \text{so that} \quad \frac{(m, n)}{N} \rightarrow (\Psi_1(\rho-\lambda), \Psi_1(\lambda)).$$

The first inequality of (4.18) follows from showing that $\forall \varepsilon > 0 \exists a > 0$ such that

$$(4.19) \quad \mathbb{P}\{I_{(k, \ell), (m+1, n+1)}^{e_2} \geq I_{(k, \ell), (m+1, n+1)} + \varepsilon\} \leq 2e^{-aN}.$$

Introduce the quenched path measure $Q_{(k,\ell),(m,n)}^{\text{NE}}$ that corresponds to the partition function in (4.11).

$$\begin{aligned} I_{(k,\ell),(m+1,n+1)}^{e_2} &= \frac{Z_{(k,\ell),(m+1,n+1)}^{\text{NE}}(t_{e_2}^* > 0)}{Q_{(k-1,\ell),(m+1,n+1)}^{\text{NE}}(t_{e_2}^* > 0) \cdot Z_{(k-1,\ell),(m+1,n+1)}^{\text{NE}}} \\ &\leq \frac{I_{(k,\ell),(m+1,n+1)}}{Q_{(k-1,\ell),(m+1,n+1)}^{\text{NE}}(t_{e_2}^* > 0)}. \end{aligned}$$

For small enough ε_N the probability in (4.19) is bounded above by the sum

$$(4.20) \quad \mathbb{P}\{I_{(k,\ell),(m+1,n+1)} \geq \frac{\varepsilon}{2\varepsilon_N}\} + \mathbb{P}\{Q_{(k-1,\ell),(m+1,n+1)}^{\text{NE}}(t_{e_1}^* > 0) \geq \varepsilon_N\}.$$

Note that the event in the Q^{NE} -probability was replaced by its complement. A sequence $0 < \varepsilon_N \searrow 0$ will be chosen below.

By (4.13) $I_{(k,\ell),(m+1,n+1)}$ has Gamma(α) distribution and so the first probability in (4.20) is bounded by $e^{-c\varepsilon/\varepsilon_N}$.

We show that the Q^{NE} -probability in (4.20) is actually a large deviation by replacing (m, n) with a direction that is characteristic for (α, ρ) . The next lemma contains the idea for replacing (m, n) .

LEMMA 4.4. *Let (\bar{m}, \bar{n}) satisfy $\bar{m} > m$ and $\ell < \bar{n} < n$. Then*

$$(4.21) \quad Q_{(k,\ell),(m,n)}^{\text{NE}}(t_{e_1}^* > 0) = Q_{(k,\ell),(\bar{m},\bar{n})}^{\text{NE}}(t_{e_1}^* > \bar{m} - m).$$

Proof. A path in $\Pi_{(k,\ell),(m,n)}$ that satisfies $t_{e_1}^* > 0$ must use one of the edges $\{(i, \bar{n}-1), (i, \bar{n})\}$, $k \leq i \leq m-1$. Otherwise it hits the east boundary first and $t_{e_1}^* = 0$. Decomposing according to this choice of edge and using definition (4.11),

$$Q_{(k,\ell),(m,n)}^{\text{NE}}(t_{e_1}^* > 0) = \sum_{i=k}^{m-1} \check{Z}_{(k,\ell),(i,\bar{n}-1)} \frac{1}{\check{\xi}_{i,\bar{n}-1}} \cdot \frac{Z_{(i,\bar{n}),(\bar{m},\bar{n})}^{\text{NE}}}{Z_{(k,\ell),(\bar{m},\bar{n})}^{\text{NE}}}.$$

By Lemma 4.3 the last ratio does not depend on (m, n) , and (m, n) can be replaced by (\bar{m}, \bar{n}) . This moves the northeast corner in definition (4.11) to (\bar{m}, \bar{n}) , as well as the reference point of $t_{e_1}^*$ in (4.14). Since the sum still runs up to $m-1$, it now represents paths in $\Pi_{(k,\ell),(\bar{m},\bar{n})}$ that hit the north boundary to the left of (m, \bar{n}) . This proves Lemma 4.4. \square

Take $(\bar{m}, \bar{n}) = (\lfloor N\Psi_1(\rho - \alpha) \rfloor + k - 1, \lfloor N\Psi_1(\alpha) \rfloor + \ell)$, essentially the characteristic direction for (α, ρ) . Since $\lambda < \alpha$ and Ψ_1 is strictly decreasing, there exists $\gamma > 0$ such that for large enough N , $\bar{m} \geq m + 1 + N\gamma$ and $\bar{n} \leq n - N\gamma$. Put $\varepsilon_N = e^{-\delta_1 \gamma N}$ for a small enough $\delta_1 > 0$. Then for large enough N ,

$$(4.22) \quad \begin{aligned} \mathbb{P}\{Q_{(k-1,\ell),(m+1,n+1)}^{\text{NE}}(t_{e_1}^* > 0) \geq \varepsilon_N\} &\leq \mathbb{P}\{Q_{(k-1,\ell),(\bar{m},\bar{n})}^{\text{NE}}(t_{e_1}^* > N\gamma) \geq e^{-\delta_1 \gamma N}\} \\ &\leq e^{-c_1 \gamma N}. \end{aligned}$$

The last inequality came from Lemma 9.2 in the appendix where we can take $\kappa_N = 1$ and $\delta \leq \gamma$. Both probabilities in (4.20) have been shown to decay exponentially in N , and consequently (4.19) holds. This completes the proof of Lemma 4.2. \square

Turning to the proof of Theorem 4.1, we begin by showing the a.s. convergence in (4.3) for a fixed sequence and fixed λ . Define ratio variables by

$$(4.23) \quad \eta_{x,(m,n)} = \frac{Z_{x,(m,n)}}{Z_{x-e_1,(m,n)}} \quad \text{and} \quad \zeta_{y,(m,n)} = \frac{Z_{y,(m,n)}}{Z_{y-e_2,(m,n)}}$$

for $x \in \mathbb{N} \times \mathbb{Z}_+$ and $y \in \mathbb{Z}_+ \times \mathbb{N}$.

PROPOSITION 4.5. *Fix $0 < \lambda < \rho$ and fix a sequence $(m, n) \rightarrow \infty$ as in (4.2). Then for all $x \in \mathbb{N} \times \mathbb{Z}_+$ and $y \in \mathbb{Z}_+ \times \mathbb{N}$ the almost sure limits*

$$(4.24) \quad \eta_x = \lim_{(m,n) \rightarrow \infty} \eta_{x,(m,n)} \quad \text{and} \quad \zeta_y = \lim_{(m,n) \rightarrow \infty} \eta_{y,(m,n)}$$

exist and have distributions $\eta_x \sim \text{Gamma}(\lambda)$ and $\zeta_y \sim \text{Gamma}(\rho - \lambda)$.

Proof. We treat the case of the η variables, the case for ζ being identical. For a while, until otherwise indicated, we are considering a fixed sequence of lattice points that satisfies $(m, n) \rightarrow \infty$ as in (4.2). To avoid extra notation we refrain from indexing the lattice points, as in (m_k, n_k) . Later we can improve the result so that the limit only depends on λ and not on the particular sequence $(m, n) \rightarrow \infty$.

We show that for $0 < s < \infty$ the distribution functions

$$(4.25) \quad G^*(s) = \mathbb{P}\left\{ \overline{\lim}_{(m,n) \rightarrow \infty} \eta_{x,(m,n)} \leq s \right\} \quad \text{and} \quad G_*(s) = \mathbb{P}\left\{ \underline{\lim}_{(m,n) \rightarrow \infty} \eta_{x,(m,n)} \leq s \right\}$$

satisfy $G^*(s) = G_*(s) = F_\lambda(s)$ where

$$F_\lambda(s) = \Gamma(\lambda)^{-1} \int_0^s t^{\lambda-1} e^{-t} dt$$

is the c.d.f. of the $\text{Gamma}(\lambda)$ distribution. Since $\underline{\lim} \eta_{x,(m,n)} \leq \overline{\lim} \eta_{x,(m,n)}$, this suffices for the conclusion. Working with the distributions allows us to use any particular construction of the processes.

Let $\{U_{i,j}\}$ be i.i.d. $\text{Uniform}(0, 1)$ random variables. For $i, j \in \mathbb{N}$ and $\alpha \in (0, \rho)$ define

$$(4.26) \quad \bar{\eta}_{i,0}^\alpha = F_\alpha^{-1}(U_{i,0}) \quad \text{and} \quad \bar{\zeta}_{0,j}^\alpha = F_{\rho-\alpha}^{-1}(U_{0,j}).$$

This gives us coupled weights $\bar{\eta}_{i,0}^\alpha \sim \text{Gamma}(\alpha)$ on the south boundary and $\bar{\zeta}_{0,j}^\alpha \sim \text{Gamma}(\rho - \alpha)$ on the west boundary of the positive quadrant. For the bulk weights take an i.i.d. collection $\{\sigma_x\}_{x \in \mathbb{N}^2}$ of $\text{Gamma}(\rho)$ weights independent of $\{U_{i,j}\}$.

As mentioned after Definition 3.1, the mutually independent initial weights $\{\sigma_{i,j}, \bar{\eta}_{i,0}^\alpha, \bar{\zeta}_{0,j}^\alpha : i, j \in \mathbb{N}\}$ can be extended to the full gamma (α, ρ) system $(\sigma, \bar{\eta}^\alpha, \bar{\zeta}^\alpha, \check{\sigma}^{[\alpha]})$. The construction preserves monotonicity of the edge weights, so that

$$(4.27) \quad \bar{\eta}_{i,j}^\alpha \leq \bar{\eta}_{i,j}^\nu \quad \text{and} \quad \bar{\zeta}_{i,j}^\alpha \geq \bar{\zeta}_{i,j}^\nu \quad \text{for } \alpha \leq \nu.$$

Bracketed superscript $[\alpha]$ reminds us that even though the variables $\{\check{\sigma}_{i,j}^{[\alpha]}\}_{i,j \geq 0}$ are i.i.d. $\text{Gamma}(\rho)$ for each $\alpha \in (0, \rho)$, they were computed from α -boundary conditions. Define partition functions

$$(4.28) \quad \check{Z}_{u,v}^{[\alpha]} = \sum_{x \in \Pi_{u,v}} \prod_{i=0}^{|v-u|_1-1} (\check{\sigma}_{x_i}^{[\alpha]})^{-1}, \quad 0 \leq u \leq v \text{ in } \mathbb{Z}^2,$$

and edge ratio weights

$$(4.29) \quad \check{I}_{x,(m,n)}^{[\alpha]} = \frac{\check{Z}_{x,(m,n)}^{[\alpha]}}{\check{Z}_{x-e_1,(m,n)}^{[\alpha]}} \quad \text{and} \quad \check{J}_{y,(m,n)}^{[\alpha]} = \frac{\check{Z}_{y,(m,n)}^{[\alpha]}}{\check{Z}_{y-e_2,(m,n)}^{[\alpha]}}.$$

For each $\alpha \in (0, \rho)$, we have equality in distribution of processes:

$$(4.30) \quad \{\check{I}_{(i+1,j),(m,n)}^{[\alpha]}, \check{J}_{(i,j+1),(m,n)}^{[\alpha]}, \check{\sigma}_{i,j}^{[\alpha]}\} \stackrel{d}{=} \{\eta_{(i+1,j),(m,n)}, \zeta_{(i,j+1),(m,n)}, w_{i,j}\}.$$

These processes are indexed by $\{(i, j) \in \mathbb{Z}_+^2, (m, n) \geq (i+1, j+1)\}$. The equality in distribution comes from identical constructions applied to i.i.d. Gamma(ρ) weights: on the left to $\check{\sigma}^{[\alpha]}$, on the right to w . Now in (4.25) we can use any process $\{\check{I}_{x,(m,n)}^{[\alpha]}\}$.

Applying Lemma 4.2 to two gamma systems $(\sigma, \bar{\eta}^{\alpha_1}, \bar{\zeta}^{\alpha_1}, \check{\sigma}^{[\alpha_1]})$ and $(\sigma, \bar{\eta}^{\alpha_2}, \bar{\zeta}^{\alpha_2}, \check{\sigma}^{[\alpha_2]})$ gives, for any $0 < \alpha_1 < \lambda < \alpha_2 < \rho$,

$$(4.31) \quad \lim_{(m,n) \rightarrow \infty} \check{I}_{(k,\ell),(m,n)}^{[\alpha_1]} \geq \bar{\eta}_{k,\ell}^{\alpha_1} \quad \text{and} \quad \overline{\lim}_{(m,n) \rightarrow \infty} \check{I}_{(k,\ell),(m,n)}^{[\alpha_2]} \leq \bar{\eta}_{k,\ell}^{\alpha_2} \quad \text{a.s.}$$

By the equality in distribution (4.30),

$$G_*(s) = \mathbb{P}\left\{ \lim_{(m,n) \rightarrow \infty} \check{I}_{(k,\ell),(m,n)}^{[\alpha_1]} \leq s \right\} \leq \mathbb{P}\{\bar{\eta}_{k,\ell}^{\alpha_1} \leq s\} = F_{\alpha_1}(s) \searrow F_\lambda(s) \quad \text{as } \alpha_1 \nearrow \lambda,$$

and

$$G^*(s) = \mathbb{P}\left\{ \overline{\lim}_{(m,n) \rightarrow \infty} \check{I}_{(k,\ell),(m,n)}^{[\alpha_2]} \leq s \right\} \geq F_{\alpha_2}(s) \nearrow F_\lambda(s) \quad \text{as } \alpha_2 \searrow \lambda.$$

This gives $F_\lambda(s) \leq G^*(s) \leq G_*(s) \leq F_\lambda(s)$ and completes the proof of Proposition 4.5. \square

Proposition 4.5 gave the a.s. convergence of ratios along a fixed sequence and for a given $\lambda \in (0, \rho)$. Next we construct a system of weights (ξ, η, ζ, w) from the limits (4.24) by defining

$$(4.32) \quad \xi_x = \eta_x + \zeta_x, \quad \text{for } x \in \mathbb{N}^2.$$

PROPOSITION 4.6. *The collection (ξ, η, ζ, w) is a gamma system of weights with parameters (λ, ρ) , that is, it satisfies Definition 3.1.*

Proof. Equations (3.1) follow from the limits (4.24) and

$$w_x = \frac{Z_{x+e_1,(m,n)} + Z_{x+e_2,(m,n)}}{Z_{x,(m,n)}}.$$

By the equality in distribution in (4.30), it also follows that the limits in (4.31) exist:

$$(4.33) \quad \check{I}_{k,\ell}^{[\alpha]} = \lim_{(m,n) \rightarrow \infty} \check{I}_{(k,\ell),(m,n)}^{[\alpha]} \quad \text{a.s.}$$

Let $0 < \alpha_1 < \lambda < \alpha_2 < \rho$. Utilizing (4.30), (4.31), and (4.33),

$$(4.34) \quad \begin{aligned} (\eta, w) &\stackrel{d}{=} (\check{I}^{[\alpha_1]}, \check{\sigma}^{[\alpha_1]}) \geq (\bar{\eta}^{\alpha_1}, \check{\sigma}^{[\alpha_1]}) \xrightarrow[\alpha_1 \nearrow \lambda]{} (\bar{\eta}^\lambda, \check{\sigma}^{[\lambda]}) \\ \text{and } (\eta, w) &\stackrel{d}{=} (\check{I}^{[\alpha_2]}, \check{\sigma}^{[\alpha_2]}) \leq (\bar{\eta}^{\alpha_2}, \check{\sigma}^{[\alpha_2]}) \xrightarrow[\alpha_2 \searrow \lambda]{} (\bar{\eta}^\lambda, \check{\sigma}^{[\lambda]}). \end{aligned}$$

The inequalities and the convergence are a.s. and coordinatewise. The convergence follows from the continuity of the definitions (4.26) in α and the continuity in the equations (3.1) that inductively define the $(\bar{\eta}^\alpha, \bar{\zeta}^\alpha, \bar{\sigma}^{[\alpha]})$ weights. The consequence is that

$$(4.35) \quad (\eta, w) \stackrel{d}{=} (\bar{\eta}^\lambda, \bar{\sigma}^{[\lambda]}).$$

Equations $\zeta_x = w_{x-e_2} - \eta_{x-e_2+e_1}$ and $\xi_x = \eta_x + \zeta_x$ map (η, w) to the full system (ξ, η, ζ, w) . The same mapping applied to the right-hand side of (4.35) recreates the system $(\sigma, \bar{\eta}^\lambda, \bar{\zeta}^\lambda, \bar{\sigma}^{[\lambda]})$, which we know to be a (λ, ρ) gamma system by its construction below (4.26). \square

Proof of Theorem 4.1. Fix a countable dense subset D of $(0, \rho)$ and $\forall \lambda \in D$ a sequence $(m, n) \rightarrow \infty$ that satisfies (4.2). By Propositions 4.5 and 4.6, we can use limits (4.3) along these particular sequences to define, almost surely, (λ, ρ) gamma systems $(\xi^\lambda, \eta^\lambda, \zeta^\lambda, w)$ for $\lambda \in D$. Monotonicity (4.4) is satisfied a.s. for $\lambda_1, \lambda_2 \in D$ by Lemma 4.2. (The point is that the \tilde{Z} partition functions in (4.6) are the same for all systems $(\xi^\lambda, \eta^\lambda, \zeta^\lambda, w)$.)

Monotonicity and known gamma distributions give also the limits in (4.5) when $\lambda \rightarrow \nu$ in D . For example, suppose $\lambda \nearrow \nu$ in D . Then $\lim_{\lambda \nearrow \nu} \eta_x^\lambda \leq \eta_x^\nu$, but both are $\text{Gamma}(\nu)$ distributed and hence coincide a.s. The limit $\xi_x^\lambda \rightarrow \xi_x^\nu$ comes from the limits of η and ζ and $\xi_x^\lambda = \eta_x^\lambda + \zeta_x^\lambda$.

Extend the weights to all $\lambda \in (0, \rho)$ by defining

$$(4.36) \quad \eta_x^\lambda = \inf\{\eta_x^\nu : \nu \in D \cap (\lambda, \rho)\} = \sup\{\eta_x^\alpha : \alpha \in D \cap (0, \lambda)\}$$

with the obvious counterpart for ζ_x^λ and then $\xi_x^\lambda = \eta_x^\lambda + \zeta_x^\lambda$. The inf and the sup in (4.36) must agree a.s. because (i) the sup is not above the inf on account of the monotonicity for $\lambda \in D$, and (ii) they are both $\text{Gamma}(\lambda)$ distributed. By the same reasoning, for $\lambda \in D$ definition (4.36) gives a.s. back the same value η_x^λ as originally constructed.

To check that the new system $(\xi^\lambda, \eta^\lambda, \zeta^\lambda, w)$ is a (λ, ρ) gamma system, fix a sequence $D \ni \alpha_i \nearrow \lambda$, and observe that equations (3.1) are preserved by limits and the correct distributions come also through the limit. Extending properties (iii) utilizes monotonicity again. Limits (4.3) of ratios for a value $\lambda \notin D$ come from Lemma 4.2.

The final point is the uniqueness in part (i). Lemma 4.2 and the limits (4.3) imply that $\eta_x^{\alpha_1} \leq \tilde{\eta}_x \leq \eta_x^{\alpha_2}$ for all $\alpha_1 < \nu < \alpha_2$. \square

5. BUSEMANN FUNCTIONS AND A VARIATIONAL CHARACTERIZATION OF THE FREE ENERGY

In this section we turn the limits of ratios of point-to-point partition functions into Busemann functions, and use these to solve a variational formula for the limiting free energy. The parts from this section needed for the sequel are definition (5.12) of the velocity $\mathbf{u}(h)$ associated to a tilt h , and the large deviation bound (5.20). The latter is needed for the proofs in Section 6.

We consider briefly general i.i.d. weights $w = (w_x)_{x \in \mathbb{Z}_+^2}$ on a probability space $(\Omega, \mathfrak{S}, \mathbb{P})$ assumed to satisfy

$$(5.1) \quad \exists \varepsilon > 0 : \quad \mathbb{E}(|\log w_0|^{2+\varepsilon}) < \infty.$$

Later we specialize back to $w_0 \sim \text{Gamma}(\rho)$. It is convenient to use exponential Boltzmann-Gibbs factors. Let $p(e_1) = p(e_2) = 1/2$ be the kernel of the background random walk X_n with expectation E and initial point $X_0 = 0$. Define the potential $g(w) = -\log w_0 + \log 2$. In this notation the point-to-point partition function (4.1) is

$$(5.2) \quad Z_{0,v} = E[e^{\sum_{k=0}^{n-1} g(T_{X_k} \omega)}, X_n = v], \quad n = |v|_1.$$

Introduce a tilted point-to-line partition function

$$(5.3) \quad Z_{0,(N)}^h = E[e^{\sum_{k=0}^{N-1} g(T_{X_k} \omega) + h \cdot X_N}], \quad h = (h_1, h_2) \in \mathbb{R}^2 \text{ and } N \in \mathbb{N}.$$

The set of limit velocities for admissible walks in \mathbb{Z}_+^2 is $\mathcal{U} = \{(u, 1-u) : 0 \leq u \leq 1\}$, with relative interior $\text{int } \mathcal{U} = \{(u, 1-u) : 0 < u < 1\}$. For each $\mathbf{u} = (u, 1-u) \in \mathcal{U}$, let $\hat{x}_n(\mathbf{u}) = (\lfloor nu \rfloor, n - \lfloor nu \rfloor)$. Define limiting point-to-point free energies

$$(5.4) \quad \Lambda_{p2p}(\mathbf{u}) = \lim_{n \rightarrow \infty} n^{-1} \log Z_{0, \hat{x}_n(\mathbf{u})}, \quad \mathbf{u} \in \mathcal{U},$$

and tilted point-to-line free energies

$$(5.5) \quad \Lambda_{p2\ell}(h) = \lim_{N \rightarrow \infty} N^{-1} \log Z_{0,(N)}^h, \quad h = (h_1, h_2) \in \mathbb{R}^2.$$

Under assumption (5.1) these limits exist \mathbb{P} -a.s., Λ_{p2p} is continuous and concave in \mathbf{u} , and $\Lambda_{p2\ell}$ is continuous and convex in h [24].

We recall two variational formulas, valid for i.i.d. weights under assumption (5.1). First, a convex duality between the free energies ([24, Rmk. 4.2], also proved below in (5.19)):

$$(5.6) \quad \Lambda_{p2p}(\mathbf{u}) = \inf_{h \in \mathbb{R}^2} \{\Lambda_{p2\ell}(h) - \mathbf{u} \cdot h\}.$$

Let \mathcal{K} denote the class of *correctors* $F : \Omega \times \{e_1, e_2\} \rightarrow \mathbb{R}$ that satisfy $F \in L^1$, $\mathbb{E}F(w, z) = 0$ for $z \in \{e_1, e_2\}$, and a cocycle property: $F(w, e_1) + F(T_{e_1}w, e_2) = F(w, e_2) + F(T_{e_2}w, e_1)$ \mathbb{P} -a.s. Then we have the variational formula [25, Thm. 2.3]

$$(5.7) \quad \Lambda_{p2\ell}(h) = \inf_{F \in \mathcal{K}} \mathbb{P}\text{-ess sup}_w \log \sum_{z \in \{e_1, e_2\}} p(z) e^{g(w) + h \cdot z + F(w, z)}.$$

We solve (5.6) and (5.7) for the log-gamma model. The next corollary turns the limits of Theorem 4.1 into Busemann functions, and states the properties needed for the development that follows. Recall the function $\theta(\mathbf{u}) \in [0, \rho]$ of (3.14), the unique parameter such that \mathbf{u} is the characteristic direction for $(\theta(\mathbf{u}), \rho)$.

COROLLARY 5.1 (Corollary of Theorem 4.1). *Assume $\{w_x\}$ are i.i.d. $\text{Gamma}(\rho)$.*

(a) *For each velocity $\mathbf{u} \in \text{int } \mathcal{U}$ and for each $x, v \in \mathbb{Z}_+^2$ the \mathbb{P} -almost sure limit*

$$(5.8) \quad B^{\mathbf{u}}(w, x) = \lim_{n \rightarrow \infty} (\log Z_{0, \hat{x}_n(\mathbf{u})+v} - \log Z_{x, \hat{x}_n(\mathbf{u})+v})$$

exists and is independent of v .

(b) *The sequences $\{B^{\mathbf{u}}(T_{ie_1}w, e_1) : i \in \mathbb{Z}_+\}$ and $\{B^{\mathbf{u}}(T_{je_2}w, e_2) : j \in \mathbb{Z}_+\}$ are i.i.d. with $e^{-B^{\mathbf{u}}(w, e_1)} \sim \text{Gamma}(\theta(\mathbf{u}))$ and $e^{-B^{\mathbf{u}}(w, e_2)} \sim \text{Gamma}(\rho - \theta(\mathbf{u}))$.*

We call $B^{\mathbf{u}}$ a Busemann function, by analogy with the Busemann functions of last-passage percolation which would be limits of differences $G_{0,\hat{x}_n(\mathbf{u})+v} - G_{x,\hat{x}_n(\mathbf{u})+v}$. Of course we are merely re-expressing limits (4.3) in the form

$$e^{-B^{\mathbf{u}}(T_x w, e_1)} = \lim_{n \rightarrow \infty} \frac{Z_{x+e_1, \hat{x}_n(\mathbf{u})+v}}{Z_{x, \hat{x}_n(\mathbf{u})+v}} = \eta_{x+e_1}^{\theta(\mathbf{u})}.$$

The admission of the perturbation v in (5.8) gives the cocycle property:

$$(5.9) \quad B^{\mathbf{u}}(w, x) + B^{\mathbf{u}}(T_x w, y) = B^{\mathbf{u}}(w, x + y).$$

As a function of $\mathbf{u} \in \text{int } \mathcal{U}$, define the tilt vector

$$(5.10) \quad h(\mathbf{u}) = (h_1(\mathbf{u}), h_2(\mathbf{u})) = - \sum_{i=1}^2 \mathbb{E}[B^{\mathbf{u}}(w, e_i)]e_i = (\Psi_0(\theta(\mathbf{u})), \Psi_0(\rho - \theta(\mathbf{u}))).$$

Note that $h(\mathbf{u})$ is not well-defined for \mathbf{u} on the axes. $\theta(\mathbf{u})$ converges to 0 (to ρ) as \mathbf{u} approaches the y -axis (x -axis). Then one of the coordinates of $h(\mathbf{u})$ approaches $-\infty$. The function

$$(5.11) \quad \mathbf{u} = (u, 1 - u) \mapsto h_1(\mathbf{u}) - h_2(\mathbf{u}) = \Psi_0(\theta(\mathbf{u})) - \Psi_0(\rho - \theta(\mathbf{u}))$$

is a continuous, strictly increasing function from $u \in (0, 1)$ onto $(-\infty, \infty)$. An inverse function to (5.10), $\mathbb{R}^2 \ni h \mapsto \mathbf{u}(h) \in \text{int } \mathcal{U}$, is given by

$$(5.12) \quad \begin{aligned} \mathbf{u} &= \mathbf{u}(h) \text{ uniquely characterized by the equation} \\ h_1 - h_2 &= \Psi_0(\theta(\mathbf{u})) - \Psi_0(\rho - \theta(\mathbf{u})). \end{aligned}$$

Note that $\mathbf{u}(h)$ is constant when h ranges along a 45 degree diagonal. If $h = 0$ there is no tilt, $\mathbf{u}(0) = (1/2, 1/2)$, and $\theta(\mathbf{u}(0)) = \rho/2$.

From these ingredients we solve (5.6).

THEOREM 5.2. *Let $\mathbf{u} = (u, 1 - u) \in \text{int } \mathcal{U}$. Tilt $h(\mathbf{u})$ kills the point-to-line free energy: $\Lambda_{p2\ell}(h(\mathbf{u})) = 0 \ \forall u \in \text{int } \mathcal{U}$. Furthermore, $h(\mathbf{u})$ minimizes in (5.6) and so*

$$(5.13) \quad \Lambda_{p2p}(\mathbf{u}) = -\mathbf{u} \cdot h(\mathbf{u}) = -u\Psi_0(\theta(\mathbf{u})) - (1 - u)\Psi_0(\rho - \theta(\mathbf{u})).$$

Define the corrector

$$(5.14) \quad F^{\mathbf{u}}(w, z) = -B^{\mathbf{u}}(w, z) - h(\mathbf{u}) \cdot z, \quad z \in \{e_1, e_2\}.$$

THEOREM 5.3. *Given $h = (h_1, h_2) \in \mathbb{R}^2$, the equation*

$$(5.15) \quad h_1(\mathbf{u}) - h_2(\mathbf{u}) = h_1 - h_2$$

determines a unique $\mathbf{u} \in \text{int } \mathcal{U}$. Then $F^{\mathbf{u}} \in \mathcal{K}$ is a minimizer in (5.7). The right-hand side of (5.7) is constant in w so the essential supremum can be dropped: \mathbb{P} -a.s.,

$$(5.16) \quad \begin{aligned} \Lambda_{p2\ell}(h) &= \log \sum_{z \in \mathcal{R}} p(z) e^{g(w) + h \cdot z + F^{\mathbf{u}}(w, z)} = -h_2(\mathbf{u}) + h_2 \\ &= -\Psi_0(\rho - \theta(\mathbf{u})) + h_2. \end{aligned}$$

Remark 5.4. Theorem 5.2 is the third proof of the explicit value of $\Lambda_{p2p}(\mathbf{u})$. This result was first derived in [27] together with fluctuation bounds. The simplest proof is in [15] where the minimization of the limit of the right-hand side of (3.11) is done with convex analysis. The value (5.16) of the tilted point-to-line free energy has not been computed before.

Remark 5.5 (Large deviations). Let us observe how the duality between tilt h and velocity \mathbf{u} in (5.6) is a standard large deviation duality. The tilted quenched path measure is

$$(5.17) \quad Q_{0,(N)}^h\{x.\} = \frac{1}{Z_{0,(N)}^h} e^{\sum_{k=0}^{N-1} g(T_{x_k}\omega) + h \cdot X_N} P\{x.\}.$$

The quenched large deviation rate function for the velocity is

$$(5.18) \quad \begin{aligned} I_h(\mathbf{v}) &= - \lim_{\delta \searrow 0} \overline{\lim}_{N \rightarrow \infty} N^{-1} \log Q_{0,(N)}^h\{|N^{-1}X_N - \mathbf{v}| \leq \delta\} \quad (\mathbb{P}\text{-a.s.}) \\ &= \Lambda_{p2\ell}(h) - h \cdot \mathbf{v} - \Lambda_{p2p}(\mathbf{v}). \end{aligned}$$

The last equality uses the continuity of Λ_{p2p} and Lemma 2.9 in [24]. The limiting logarithmic moment generating function is

$$\Lambda_{Q,h}(a) = \lim_{N \rightarrow \infty} N^{-1} \log E^{Q_{0,(N)}^h} [e^{a \cdot X_N}] = \Lambda_{p2\ell}(h + a) - \Lambda_{p2\ell}(h) \quad \mathbb{P}\text{-a.s.}$$

By Varadhan's theorem these are convex duals of each other:

$$(5.19) \quad I_h(\mathbf{v}) = \sup_{a \in \mathbb{R}^2} \{a \cdot \mathbf{v} - \Lambda_{Q,h}(a)\}$$

which is the same as (5.6). For the next section we need the minimizer of I_h . By (3.14), (5.13) and calculus, I_h is uniquely minimized by $\mathbf{u}(h)$ defined by (5.12). Consequently the walk converges exponentially fast: for $\delta > 0$,

$$(5.20) \quad \overline{\lim}_{N \rightarrow \infty} N^{-1} \log Q_{0,(N)}^h\{|N^{-1}X_N - \mathbf{u}(h)| \geq \delta\} < 0 \quad \mathbb{P}\text{-a.s.}$$

The function $\Lambda_{p2p}(\mathbf{u})$ extends naturally to all of \mathbb{R}_+^2 by homogeneity: $\Lambda_{p2p}(c\mathbf{u}) = c\Lambda_{p2p}(\mathbf{u})$. Part of the duality setting is that the mean of the Busemann function gives the gradient $\nabla_{\mathbf{u}} \Lambda_{p2p}(\mathbf{u}) = -h(\mathbf{u})$.

The remainder of this section proves the theorems.

Proof of Theorem 5.2. That $F^{\mathbf{u}}$ is a corrector is clear by (5.9). Let

$$f^{\mathbf{u}}(w, x) = \sum_{i=0}^{m-1} F^{\mathbf{u}}(T_{x_i}w, x_{i+1} - x_i) = -B^{\mathbf{u}}(w, x) - h(\mathbf{u}) \cdot x$$

be the path integral of F . The admissible path $\{x_i\}_{i=0}^m$ above satisfies $x_0 = 0$ and $x_m = x$, and the cocycle property implies that $f^{\mathbf{u}}$ depends on the path only through the endpoint x . Corollary 5.1(b) verifies exactly the sufficient condition (9.11) for (9.10). From Theorem 9.3 in the appendix,

$$(5.21) \quad \max_{x \in \mathbb{Z}_+^d : |x|_1 = n} \frac{|f^{\mathbf{u}}(w, x)|}{n} \rightarrow 0 \quad \text{a.s.}$$

This ergodic property slips $f^{\mathbf{u}}(w, X_n)$ into the exponent in the free energy limit, and shows that tilt $h(\mathbf{u})$ kills the point-to-line free energy:

$$\begin{aligned}
 \Lambda_{p2\ell}(h(\mathbf{u})) &= \lim_{n \rightarrow \infty} n^{-1} \log E[e^{\sum_{k=0}^{n-1} g(T_{X_k} w) + h(\mathbf{u}) \cdot X_n}] \\
 &= \lim_{n \rightarrow \infty} n^{-1} \log E[e^{\sum_{k=0}^{n-1} g(T_{X_k} w) + h(\mathbf{u}) \cdot X_n + f^{\mathbf{u}}(w, X_n)}] \\
 &= \lim_{n \rightarrow \infty} n^{-1} \log E[e^{\sum_{k=0}^{n-1} (g(T_{X_k} w) + h(\mathbf{u}) \cdot Z_{k+1} + F^{\mathbf{u}}(T_{X_k} w, Z_{k+1}))}] \\
 &= 0.
 \end{aligned}
 \tag{5.22}$$

The last equality comes from

$$\begin{aligned}
 \sum_z p(z) e^{g(w) + h(\mathbf{u}) \cdot z + F^{\mathbf{u}}(w, z)} &= \sum_z p(z) e^{g(w) - B^{\mathbf{u}}(w, z)} \\
 &= \lim_{n \rightarrow \infty} \frac{\sum_z p(z) e^{g(w)} Z_{z, \hat{x}_n(\mathbf{u})}}{Z_{0, \hat{x}_n(\mathbf{u})}} = \lim_{n \rightarrow \infty} \frac{Z_{0, \hat{x}_n(\mathbf{u})}}{Z_{0, \hat{x}_n(\mathbf{u})}} = 1.
 \end{aligned}
 \tag{5.23}$$

Fix $\mathbf{u} \in \mathcal{U}$. Since $|X_n|_1 = n$, the expression on the right-hand side of (5.6) satisfies

$$\Lambda_{p2\ell}(h) - \mathbf{u} \cdot h = \Lambda_{p2\ell}(h_1 - h_2, 0) - \mathbf{u} \cdot (h_1 - h_2, 0)$$

and so, as a function of h , is constant along 45 degree diagonals. So the minimization needs one h point from each diagonal, which is what parameterization $h(\mathbf{v})$ of (5.10) achieves by virtue of the bijection (5.11). The upshot is that

$$\begin{aligned}
 \Lambda_{p2\ell}(\mathbf{u}) &= \inf_{\mathbf{v} \in \text{int } \mathcal{U}} \{ \Lambda_{p2\ell}(h(\mathbf{v})) - h(\mathbf{v}) \cdot \mathbf{u} \} \\
 &= \inf_{\mathbf{v} \in \text{int } \mathcal{U}} \{ -h(\mathbf{v}) \cdot \mathbf{u} \} = -h(\mathbf{u}) \cdot \mathbf{u}.
 \end{aligned}
 \tag{5.24}$$

The last step is calculus: from the explicit formula (5.10), $h(\mathbf{v}) \cdot \mathbf{u}$ is uniquely maximized at $\mathbf{v} = \mathbf{u}$. This completes the proof of Theorem 5.2. \square

Proof of Theorem 5.3. Since $|X_n|_1 = n$ and by (5.22),

$$\begin{aligned}
 \Lambda_{p2\ell}(h) &= \lim_{n \rightarrow \infty} n^{-1} \log E[e^{\sum_{k=0}^{n-1} g(T_{X_k} w) + h \cdot X_n}] \\
 &= \lim_{n \rightarrow \infty} n^{-1} \log E[e^{\sum_{k=0}^{n-1} g(T_{X_k} w) + h(\mathbf{u}) \cdot X_n}] - h_2(\mathbf{u}) + h_2 = -h_2(\mathbf{u}) + h_2.
 \end{aligned}$$

On the other hand, by (5.23),

$$\begin{aligned}
 \log \sum_z p(z) e^{g(w) + h \cdot z + F^{\mathbf{u}}(w, z)} &= \log \sum_z p(z) e^{g(w) + h(\mathbf{u}) \cdot z + F^{\mathbf{u}}(w, z)} - h_2(\mathbf{u}) + h_2 \\
 &= -h_2(\mathbf{u}) + h_2.
 \end{aligned}
 \tag{5.25}$$

\square

6. LIMITS OF RATIOS OF POINT-TO-LINE PARTITION FUNCTIONS

Armed with the limits of Theorem 4.1 and the large deviation bound of Remark 5.5, we prove convergence of ratios of tilted point-to-line partition functions. With the tilt parameter $h = (h_1, h_2) \in \mathbb{R}^2$ and $Z_{u,v}$ defined as in (4.1), let

$$Z_{u,(N)}^h = \sum_{v: |v|_1 = N} e^{h \cdot (v-u)} Z_{u,v} \quad \text{for } N \in \mathbb{N} \text{ and } |u|_1 \leq N.$$

This is the same as (5.3) with a general initial point. Recall definition (5.12) that associates a velocity $\mathbf{u}(h) = (u(h), 1-u(h))$ to a tilt h , and definition (3.14) that associates a parameter $\theta(\mathbf{v})$ to a velocity \mathbf{v} .

THEOREM 6.1. *Fix $0 < \rho < \infty$ and let i.i.d. Gamma(ρ) weights $\{w_x\}_{x \in \mathbb{Z}_+^2}$ be given. For $\lambda \in (0, \rho)$, let $(\xi^\lambda, \eta^\lambda, \zeta^\lambda, w)$ be the gamma system constructed in Theorem 4.1. Then for $h = (h_1, h_2) \in \mathbb{R}^2$, $x \in \mathbb{N} \times \mathbb{Z}_+$, $y \in \mathbb{Z}_+ \times \mathbb{N}$, \mathbb{P} -a.s.,*

$$(6.1) \quad \lim_{N \rightarrow \infty} \frac{Z_{x,(N)}^h}{e^{-h_1} Z_{x-e_1,(N)}^h} = \eta_x^{\theta(\mathbf{u}(h))} \quad \text{and} \quad \lim_{N \rightarrow \infty} \frac{Z_{x,(N)}^h}{e^{-h_2} Z_{x-e_2,(N)}^h} = \zeta_x^{\theta(\mathbf{u}(h))}.$$

In other words, the limit of ratios of point-to-line partition functions tilted by h is equal to the limit of ratios of point-to-point partition functions in the direction $\mathbf{u}(h)$:

$$(6.2) \quad \begin{aligned} \lim_{N \rightarrow \infty} \frac{Z_{x,(N)}^h}{e^{-h_1} Z_{x-e_1,(N)}^h} &= \lim_{(m,n) \rightarrow \infty} \frac{Z_{x,(m,n)}}{Z_{x-e_1,(m,n)}} \\ \text{and} \quad \lim_{N \rightarrow \infty} \frac{Z_{x,(N)}^h}{e^{-h_2} Z_{x-e_2,(N)}^h} &= \lim_{(m,n) \rightarrow \infty} \frac{Z_{x,(m,n)}}{Z_{x-e_2,(m,n)}}. \end{aligned}$$

provided $m/n \rightarrow u(h)/(1-u(h))$. We see the duality between tilt and velocity from Remark 5.5 again.

In Section 7 the limits of ratios from Theorems 4.1 and 6.1 give convergence of polymer measures to random walk in a correlated random environment. The remainder of this section proves Theorem 6.1.

Proof of Theorem 6.1. We prove (6.1) for the horizontal ratios (first limit). Begin with a lower bound and let $\delta_0 > 0$.

$$\begin{aligned} \frac{Z_{x,(N)}^h}{e^{-h_1} Z_{x-e_1,(N)}^h} &= \sum_{v: |v|_1 = N} \frac{e^{h \cdot (v-x)} Z_{x-e_1,v}}{e^{-h_1} Z_{x-e_1,(N)}^h} \cdot \frac{Z_{x,v}}{Z_{x-e_1,v}} \\ &= \sum_{v: |v|_1 = N} Q_{x-e_1,(N)}^h \{X_{N-|x|_1+1} = v\} \frac{Z_{x,v}}{Z_{x-e_1,v}} \\ &\geq \sum_{m: |m-Nu(h)| < N\delta_0} Q_{x-e_1,(N)}^h \{X_{N-|x|_1+1} = (m, N-m)\} \frac{Z_{x,(m,N-m)}}{Z_{x-e_1,(m,N-m)}}. \end{aligned}$$

Above we introduced a tilted quenched point-to-line polymer measure

$$(6.3) \quad Q_{y,(N)}^h\{x,\} = \frac{1}{Z_{y,(N)}^h} e^{h \cdot (x_{N-|y|_1} - y)} \prod_{i=0}^{N-|y|_1-1} w_{x_i}^{-1}$$

for paths x_\cdot from $x_0 = y$ to the line $|x_{N-|y|_1}|_1 = N$.

Apply construction (4.11) to the gamma system $(\xi^\lambda, \eta^\lambda, \zeta^\lambda, w)$ to define partition functions Z^λ and associated polymer measures Q^λ with northern boundary weights $\{\eta_{i,N-m+1}^\lambda\}_{1 \leq i \leq m+1}$ and eastern boundary weights $\{\zeta_{m+1,j}^\lambda\}_{1 \leq j \leq N-m+1}$. Recall the dual exit points (4.14)–(4.15). By an application of Lemma 9.1 (to the reversed rectangle),

$$\begin{aligned} \frac{Z_{x,(m,N-m)}}{Z_{x-e_1,(m,N-m)}} &\geq \frac{Z_{x,(m+1,N-m+1)}^\lambda(t_{e_1}^* > 0)}{Z_{x-e_1,(m+1,N-m+1)}^\lambda(t_{e_1}^* > 0)} \\ &\geq Q_{x,(m+1,N-m+1)}^\lambda\{t_{e_1}^* > 0\} \frac{Z_{x,(m+1,N-m+1)}^\lambda}{Z_{x-e_1,(m+1,N-m+1)}^\lambda} \\ &= Q_{x,(m+1,N-m+1)}^\lambda\{t_{e_1}^* > 0\} \eta_x^\lambda. \end{aligned}$$

The last equality came from Lemma 4.3. Note the notational distinction: $Q_{y,(N)}^h$ is the tilted point-to-line polymer measure, while $Q_{x,y}^\lambda$ is the point-to-point polymer measure with boundary parameter λ .

We have the lower bound

$$(6.4) \quad \frac{Z_{x,(N)}^h}{e^{-h_1} Z_{x-e_1,(N)}^h} \geq \sum_{m: |m-Nu(h)| < N\delta_0} Q_{x-e_1,(N)}^h\{X_{N-|x|_1+1} = (m, N-m)\} \\ \times Q_{x,(m+1,N-m+1)}^\lambda\{t_{e_1}^* > 0\} \eta_x^\lambda.$$

Let $0 < \lambda < \theta(\mathbf{u}(h))$. Define parameter $M \nearrow \infty$ by $N(1-u(h)) = M\Psi_1(\theta(\mathbf{u}(h)))$. Let $(\bar{m}, \bar{n}) = x + (\lfloor M\Psi_1(\rho - \lambda) \rfloor, \lfloor M\Psi_1(\lambda) \rfloor)$, a velocity essentially characteristic for (λ, ρ) . As m varies in the sum on the left side of (6.4), let $(m_1, n_1) = (m+1, N-m+1)$. Since Ψ_1 is strictly decreasing, if we fix $\delta_0 > 0$ small enough, there exists $\varepsilon_0 > 0$ such that, for large enough N ,

$$\begin{aligned} \bar{n} - n_1 &\geq M\Psi_1(\lambda) - M\Psi_1(\theta(\mathbf{u}(h))) - N\delta_0 - 2 \geq M\varepsilon_0 \\ \text{and } m_1 - \bar{m} &\geq M\varepsilon_0. \end{aligned}$$

Following the idea of Lemma 4.4 and (4.22),

$$\begin{aligned} \mathbb{P}[Q_{x,(m+1,N-m+1)}^\lambda\{t_{e_2}^* > 0\} > e^{-\delta_1\varepsilon_0 M}] &\leq \mathbb{P}[Q_{x,(\bar{m},\bar{n})}^\lambda\{t_{e_2}^* > M\varepsilon_0\} > e^{-\delta_1\varepsilon_0 M}] \\ &\leq e^{-c_1\varepsilon_0 M}. \end{aligned}$$

Since there are $O(N)$ m -values, Borel-Cantelli and (6.4) give, for large enough n ,

$$\frac{Z_{x,(N)}^h}{e^{-h_1} Z_{x-e_1,(N)}^h} \geq \eta_x^\lambda (1 - e^{-\delta_1\varepsilon_0 M}) Q_{x-e_1,(N)}^h\{|X_{N-|x|_1+1} - N\mathbf{u}(h)| < N\delta_0\}.$$

By the quenched LDP (5.20) for the point-to-line measure, the last probability tends to 1. Thus we obtain the lower bound:

$$\lim_{N \rightarrow \infty} \frac{Z_{x,(N)}}{e^{-h_1} Z_{x-e_1,(N)}} \geq \eta_x^\lambda \nearrow \eta_x^{\theta(\mathbf{u}(h))} \quad \text{as we let } \lambda \nearrow \theta(\mathbf{u}(h)).$$

For the upper bound we first bound summands away from the concentration point of the quenched measure.

$$\begin{aligned} & \sum_{m: |m - Nu(h)| \geq N\delta_0} \frac{e^{h \cdot ((m, N-m) - x)} Z_{x,(m, N-m)}}{e^{-h_1} Z_{x-e_1,(N)}^h} \\ & \leq w_{x-e_1} \sum_{m: |m - Nu(h)| \geq N\delta_0} \frac{e^{h \cdot ((m, N-m) - x)} Z_{x,(m, N-m)}}{Z_{x,(N)}^h} \\ & \leq w_{x-e_1} Q_{x,(N)}^h \{ |X_{N-|x|_1} - N\mathbf{u}(h)| \geq N\delta_0 \} \rightarrow 0. \end{aligned}$$

For the remaining fractions we develop an upper bound.

$$\begin{aligned} \frac{Z_{x,(m, N-m)}}{Z_{x-e_1,(m, N-m)}} & \leq \frac{Z_{x,(m+1, N-m+1)}^\lambda (t_{e_2}^* > 0)}{Z_{x-e_1,(m+1, N-m+1)}^\lambda (t_{e_2}^* > 0)} \\ & \leq \frac{1}{Q_{x-e_1,(m+1, N-m+1)}^\lambda \{t_{e_2}^* > 0\}} \cdot \frac{Z_{x,(m+1, N-m+1)}^\lambda}{Z_{x-e_1,(m+1, N-m+1)}^\lambda} \\ & = \frac{\eta_x^\lambda}{Q_{x-e_1,(m+1, N-m+1)}^\lambda \{t_{e_2}^* > 0\}}. \end{aligned}$$

Combining these,

$$\begin{aligned} \frac{Z_{x,(N)}^h}{e^{-h_1} Z_{x-e_1,(N)}^h} & \leq \sum_{m: |m - Nu(h)| < N\delta_0} Q_{x-e_1,(N)}^h \{X_{N-|x|_1+1} = (m, N-m)\} \\ & \quad \times \frac{\eta_x^\lambda}{1 - Q_{x-e_1,(m+1, N-m+1)}^\lambda \{t_{e_1}^* > 0\}} + o(1) \end{aligned}$$

where the $o(1)$ term tends to zero \mathbb{P} -a.s. Proceed as for the lower bound, this time choosing $\theta(\mathbf{u}(h)) < \lambda < \rho$ to show that the Q^λ -probability above vanishes exponentially fast. This completes the proof of Theorem 6.1. \square

7. LIMITS OF PATH MEASURES

As in Section 4, fix $\rho \in (0, \infty)$ and assume that i.i.d. $\text{Gamma}(\rho)$ weights $w = \{w_x : x \in \mathbb{Z}_+^2\}$ are given on a probability space $(\Omega, \mathfrak{S}, \mathbb{P})$. Let $Z_{u,v}$ be the point-to-point partition functions defined in (4.1), with associated quenched polymer measures

$$(7.1) \quad Q_{u,v}\{x\cdot\} = \frac{1}{Z_{u,v}} \prod_{i=0}^{|v-u|_1-1} w_{x_i}^{-1}, \quad x\cdot \in \Pi_{u,v}.$$

Let point-to-line polymer measures be defined as before in (5.17) or (6.3).

For $\lambda \in (0, \rho)$, let $(\xi^\lambda, \eta^\lambda, \zeta^\lambda, w)$ denote the gamma system of weights constructed in Theorem 4.1. In this environment, define RWRE transitions on \mathbb{Z}_+^2 by

$$(7.2) \quad \pi^{w,\lambda}(x, x + e_1) = \frac{\eta_{x+e_1}^\lambda}{\eta_{x+e_1}^\lambda + \zeta_{x+e_2}^\lambda} \quad \text{and} \quad \pi^{w,\lambda}(x, x + e_2) = \frac{\zeta_{x+e_2}^\lambda}{\eta_{x+e_1}^\lambda + \zeta_{x+e_2}^\lambda}.$$

Let $P^{w,\lambda}$ be the quenched path measure of the RWRE started at 0. It is characterized by the initial point and transition

$$(7.3) \quad P^{w,\lambda}(X_0 = 0) = 1, \quad P^{w,\lambda}(X_{k+1} = y | X_k = x) = \pi^{w,\lambda}(x, y).$$

We wrote $P^{w,\lambda}$ instead of $P^{\omega,\lambda}$ because the quenched distribution is a function of the weights w , through the limits (4.3) that appear on the right in (7.2). In other words, the probability space has not been artificially augmented with the variables that appear in definition (2.16): everything comes from the single i.i.d. collection w .

Let $Z_{u,v}^\lambda$ denote partition functions defined by recipe (3.10) in gamma system $(\xi^\lambda, \eta^\lambda, \zeta^\lambda, w)$. Adapt the notation from (3.3) in the form

$$(7.4) \quad \tau_{x,x+z}^\lambda = \begin{cases} \eta_{x+e_1}^\lambda, & z = e_1 \\ \zeta_{x+e_2}^\lambda, & z = e_2. \end{cases}$$

Then we can rewrite transition (7.2) as

$$(7.5) \quad \pi^{w,\lambda}(x, x + z) = \frac{\tau_{x,x+z}^\lambda}{\tau_{x,x+e_1}^\lambda + \tau_{x,x+e_2}^\lambda} = \frac{(Z_{0,x+z}^\lambda)^{-1}}{(Z_{0,x+e_1}^\lambda)^{-1} + (Z_{0,x+e_2}^\lambda)^{-1}}, \quad z \in \{e_1, e_2\}.$$

In other words, this RWRE is of the competition interface type defined by (2.11) in Lemma 2.2. The next theorem shows that these walks are the limits of the polymer measures on long paths, both point-to-point and point-to-line.

THEOREM 7.1. *The following weak limits of probability measures on the path space $(\mathbb{Z}_+^2)^{\mathbb{Z}_+}$ happen for \mathbb{P} -a.e. w .*

(i) *Let $0 < \lambda < \rho$ and suppose $(m, n) \rightarrow \infty$ in the characteristic direction of parameters (λ, ρ) as defined in (4.2). Then $Q_{0,(m,n)}$ converges to $P^{w,\lambda}$.*

(ii) *Let $h \in \mathbb{R}^2$. Then as $N \rightarrow \infty$ the tilted point-to-line measure $Q_{0,(N)}^h$ converges to $P^{w,\theta(u(h))}$.*

Proof. Fix a finite path $x_{0,M}$ with $x_0 = 0$. Then $(m, n) \geq x_M$ for large enough (m, n) , and

$$(7.6) \quad \begin{aligned} Q_{0,(m,n)}\{X_{0,M} = x_{0,M}\} &= \frac{Z_{x_M,(m,n)}}{Z_{0,(m,n)}} \prod_{i=0}^{M-1} w_{x_i}^{-1} \xrightarrow{(m,n) \rightarrow \infty} \prod_{i=0}^{M-1} \frac{\tau_{x_i, x_{i+1}}^\lambda}{w_{x_i}} \\ &= \prod_{i=0}^{M-1} \pi^{w,\lambda}(x_i, x_{i+1}) = P^{w,\lambda}\{X_{0,M} = x_{0,M}\}. \end{aligned}$$

We applied limits (4.3) and used property $w_x = \eta_{x+e_1}^\lambda + \zeta_{x+e_2}^\lambda$ of the gamma system $(\xi^\lambda, \eta^\lambda, \zeta^\lambda, w)$ from Theorem 4.1. There are countably many finite paths and these determine weak convergence on the path space. Hence \mathbb{P} -a.s. limits (7.6) give claim (i).

The proof of (ii) is the same with limits (6.1) instead. \square

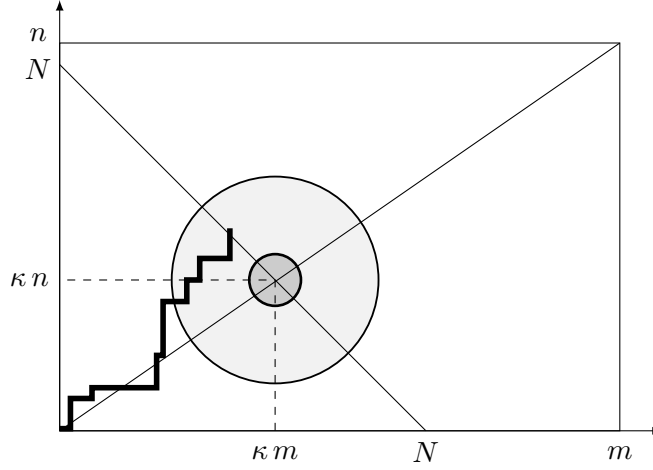


FIGURE 3. Illustration of the proof of Theorem 7.2. The thickset RWRE path avoids the disk of radius $\delta N^{2/3}$ (dark grey small disk) but enters the disk of radius $bN^{2/3}$ (light grey large disk) centered at $(\kappa m, \kappa n) = N\mathbf{u}_{\lambda, \rho}$.

The RWRE $P^{w, \lambda}$ has the fluctuation exponent of the 1+1 dimensional KPZ (Kardar-Parisi-Zhang) universality class: under the averaged distribution, at time n , the typical fluctuation away from the characteristic velocity of (λ, ρ) is of size $n^{2/3}$. The reason is that the RWRE is close to a polymer and we can apply fluctuation results for the shift-invariant log-gamma polymer. Below \mathbb{E} denotes expectation over the weights w . Recall the characteristic velocity $\mathbf{u}_{\lambda, \rho}$ from (3.13).

THEOREM 7.2. *There exist constants $C_1, C_2 < \infty$ such that for $N \in \mathbb{N}$ and $b \geq C_1$,*

$$(7.7) \quad \mathbb{E}P^{w, \lambda}\{|X_N - N\mathbf{u}_{\lambda, \rho}| \geq bN^{2/3}\} \leq C_2b^{-3}.$$

Given $\varepsilon > 0$, there exists $\delta > 0$ such that

$$(7.8) \quad \lim_{N \rightarrow \infty} \mathbb{E}P^{w, \lambda}\{|X_N - N\mathbf{u}_{\lambda, \rho}| \leq \delta N^{2/3}\} \leq \varepsilon.$$

Proof. For each N let $(m, n) = (\lfloor cN\Psi_1(\rho - \lambda) \rfloor, \lfloor cN\Psi_1(\lambda) \rfloor)$ where $c > 0$ is fixed large enough so that $m \wedge n > 2N$. Define $0 < \kappa < 1$ by $\kappa^{-1} = c(\Psi_1(\rho - \lambda) + \Psi_1(\lambda))$. Then up to errors from integer parts $(\kappa m, \kappa n) = N\mathbf{u}_{\lambda, \rho}$. (See Figure 3.)

Fix (m, n) . We couple the RWRE $P^{w, \lambda}$ with the polymer that obeys the quenched distribution $Q_{0, (m, n)}^{\lambda, \text{NE}}$ defined by applying the construction (4.11) to the gamma system $(\xi^\lambda, \eta^\lambda, \zeta^\lambda, w)$. In other words, the boundary weights η^λ and ζ^λ are on the north and east, the bulk weights come from w , and the distribution of the weights is described by (3.7), with w taking on the role of $\check{\xi}$. This is the stationary log-gamma polymer to which results from [27] apply.

Define the path $\check{X}_\bullet \in \Pi_{0, (m, n)}$ by letting it follow the RWRE until it hits either the north or the east boundary of the rectangle $\{0, \dots, m\} \times \{0, \dots, n\}$, and then follow the boundary to (m, n) . The next calculation shows that the quenched distribution of \check{X}_\bullet is $Q_{0, (m, n)}^{\lambda, \text{NE}}$. Let

$x_\bullet \in \Pi_{0,(m,n)}$. To be concrete, let $0 \leq k < m$ and suppose x_\bullet hits the north boundary at $x_{k+n} = (k, n)$.

$$\begin{aligned} P^{w,\lambda}(\check{X}_\bullet = x_\bullet) &= \prod_{j=0}^{k+n-1} \frac{\tau_{x_j, x_{j+1}}^\lambda}{w_{x_j}} = \frac{1}{Z_{0,(k,n)}^\lambda} \prod_{j=0}^{k+n-1} w_{x_j}^{-1} \\ &= \frac{1}{Z_{0,(m,n)}^\lambda} \prod_{j=0}^{k+n-1} w_{x_j}^{-1} \cdot \prod_{i=k+1}^m (\eta_{i,n}^\lambda)^{-1} = \frac{1}{Z_{0,(m,n)}^{\lambda^{\text{NE}}}} \prod_{j=0}^{k+n-1} w_{x_j}^{-1} \cdot \prod_{i=k+1}^m (\eta_{i,n}^\lambda)^{-1} \\ &= Q_{0,(m,n)}^{\lambda^{\text{NE}}} \{x_\bullet\}. \end{aligned}$$

The last equality is the definition of $Q_{0,(m,n)}^{\lambda^{\text{NE}}} \{x_\bullet\}$. The equality $Z_{0,(m,n)}^{\lambda^{\text{NE}}} = Z_{0,(m,n)}^\lambda$ comes by applying Lemma 4.3 to a telescoping product of ratio weights.

With c large enough, the boundary does not interfere with behavior around $(\kappa m, \kappa n) = N\mathbf{u}_{\lambda,\rho}$. In (7.7)–(7.8) we can replace $\mathbb{E}P^{w,\lambda}\{\cdot\}$ with $\mathbb{E}Q_{0,(m,n)}^{\lambda^{\text{NE}}}\{\cdot\}$. The result follows from Theorem 2.3 of [27], after a harmless reversal of the lattice rectangle to account for the difference that in [27, Thm. 2.3] the boundary weights are on the south and west. \square

8. THE LOG-GAMMA POLYMER RANDOM WALK IN RANDOM ENVIRONMENT

In the previous section we saw that the limits of log-gamma polymer measures are polymer RWREs with transition (2.17), where the weights come from a gamma system with some parameters (λ, ρ) . In this section we identify a stationary, ergodic probability distribution for the environment process of a polymer RWRE. We expect this stationary Markov chain to be the limit of the environment process when its initial distribution is an appropriate gamma system (Remark 8.3 below).

The process of the environment as seen from the particle is

$$T_{X_n}\omega = (\xi_{X_n+\mathbb{N}^2}, \eta_{X_n+\mathbb{N}\times\mathbb{Z}_+}, \zeta_{X_n+\mathbb{Z}_+\times\mathbb{N}}, \check{\xi}_{X_n+\mathbb{Z}_+^2}).$$

The state space of this process is the space Ω_{NE} of weight configurations $\omega = (\xi, \eta, \zeta, \check{\xi})$ that satisfy NE induction, as defined in Definition 2.3 and (2.16).

Let $0 < \alpha, \beta < \infty$ and $\rho = \alpha + \beta + 1$. Define probability distribution $\mu^{\alpha,\beta}$ on the space Ω_{NE} as follows: let the variables $(\eta_{\mathbb{N}e_1}, \zeta_{\mathbb{N}e_2}, \xi_{\mathbb{N}^2})$ be mutually independent with marginal distributions

$$(8.1) \quad \eta_{i,0} \sim \text{Gamma}(\alpha), \zeta_{0,j} \sim \text{Gamma}(\beta), \xi_{i,j} \sim \text{Gamma}(\rho), \quad i, j \in \mathbb{N}.$$

The remaining variables $\{\eta_x, \zeta_x, \check{\xi}_{x-e_1-e_2} : x \in \mathbb{N}^2\}$ are then defined by north-east induction (2.12)–(2.13).

A few more notational items. G_α denotes a $\text{Gamma}(\alpha)$ random variable and \mathbf{E} generic expectation. Let P denote the distribution of the random walk on \mathbb{Z}_+^2 that starts at 0 and has step distribution

$$p(e_1) = \frac{\alpha}{\alpha + \beta} = 1 - p(e_2).$$

Let us call this the $(\frac{\alpha}{\alpha+\beta}, \frac{\beta}{\alpha+\beta})$ random walk. An admissible path is denoted by $x_{0,n} = (x_0, x_1, \dots, x_n)$ with $x_0 = 0$ and steps $z_k = x_k - x_{k-1} \in \{e_1, e_2\}$.

The Burke property is not valid for $\mu^{\alpha,\beta}$ because $\rho \neq \alpha + \beta$, so under $\mu^{\alpha,\beta}$ the weights do not form a gamma system (Definition 3.1). However, the $\check{\xi}$ weights still turn out to have a tractable distribution which we record in the next proposition.

PROPOSITION 8.1. *Under $\mu^{\alpha,\beta}$, the marginal distribution of $\{\check{\xi}_x\}_{x \in \mathbb{Z}_+^2}$ is given as follows. Let $\{h_x\}_{x \in \mathbb{Z}_+^2}$ be arbitrary bounded Borel functions on \mathbb{R}_+ . Then for $n \in \mathbb{N}$,*

$$(8.2) \quad E^{\mu^{\alpha,\beta}} \left[\prod_{x \in \mathbb{Z}_+^2: |x|_1 \leq n} h_x(\check{\xi}_x) \right] = \sum_{x_{1,n} \in (\mathbb{Z}_+^2)^n} P(X_{0,n} = x_{0,n}) \prod_{k=0}^n \mathbf{E} h_{x_k}(G_{\alpha+\beta}) \\ \times \prod_{|y|_1 \leq n: y \notin \{x_{0,n}\}} \mathbf{E} h_y(G_{\alpha+\beta+1}).$$

In other words, the distribution of the $\check{\xi}$ weights is constructed as follows: run the $(\frac{\alpha}{\alpha+\beta}, \frac{\beta}{\alpha+\beta})$ random walk, put independent $\text{Gamma}(\alpha + \beta)$ variables on the path, independent $\text{Gamma}(\alpha + \beta + 1)$ variables off the path, and average over the walks.

THEOREM 8.2. *Let the environment ω have initial distribution $\mu^{\alpha,\beta}$ on the space Ω_{NE} of (2.16), and let the walk X_n obey transitions (2.17).*

- (a) *The environment process $T_{X_n}\omega$ is a stationary ergodic Markov chain with state space Ω_{NE} .*
- (b) *The averaged distribution of the walk X_n is the homogeneous $(\frac{\alpha}{\alpha+\beta}, \frac{\beta}{\alpha+\beta})$ random walk.*

Note the contrast in the behavior of the walk X_n . According to Theorem 7.2, when the environment has the distribution of a gamma system of weights, the averaged walk has fluctuations of order $n^{2/3}$. By part (b) above, when the environment has the $\mu^{\alpha,\beta}$ distribution, the averaged walk is diffusive.

Remark 8.3 (Simulations). Suppose the environment process starts from a gamma system with parameters (λ, ρ) , with $\rho > 1$. Simulations suggest that then $T_{X_n}\omega$ converges to $\mu^{\alpha,\beta}$ such that $\alpha + \beta = \rho - 1$ and $(\frac{\alpha}{\alpha+\beta}, \frac{\beta}{\alpha+\beta}) = \mathbf{u}_{\lambda,\rho}$, the characteristic direction (3.13) of the original setting.

Under the environment distribution $\mu^{\alpha,\beta}$, the averaged distribution of the walk X_n is the diffusive $(\frac{\alpha}{\alpha+\beta}, \frac{\beta}{\alpha+\beta})$ random walk. Simulations suggest that under its quenched distribution the walk localizes, with a positive fraction of overlap between two independent walks in the same environment.

Remark 8.4. We can look at the environment as seen from the walk with a more general boundary, instead of simply the axes. Let $\sigma = \{y_j\}_{j \in \mathbb{Z}}$ be a down-right path in \mathbb{Z}^2 that goes through e_2 , 0 and e_1 . That is, $y_{-1} = e_2$, $y_0 = 0$, $y_1 = e_1$ and $y_i - y_{i-1} \in \{e_1, -e_2\}$. Let $\mathcal{J} = \{x : \exists k \in \mathbb{N} : x - (k, k) \in \sigma\}$ be the lattice strictly to the northeast of σ . Weights assigned to this setting are such that $\{\xi_x : x \in \mathcal{J}\}$ are i.i.d. $\text{Gamma}(\rho)$. On the path edge

weights have different recipes to the northwest and southeast of the origin:

horizontal edge northwest of 0: $i < 0, y_i - y_{i-1} = e_1 : \eta_{y_i} \sim \text{Gamma}(\alpha + 1)$

vertical edge northwest of 0: $i \leq 0, y_i - y_{i-1} = -e_2 : \zeta_{y_{i-1}} \sim \text{Gamma}(\beta)$

horizontal edge southeast of 0: $i \geq 1, y_i - y_{i-1} = e_1 : \eta_{y_i} \sim \text{Gamma}(\alpha)$

vertical edge southeast of 0: $i > 1, y_i - y_{i-1} = -e_2 : \zeta_{y_{i-1}} \sim \text{Gamma}(\beta + 1)$

These weights are stationary as we look at the system centered at X_n . The proof goes along the same lines as given below.

Remark 8.5 (A degenerate limit and an invariant distribution as seen from a last-passage competition interface). The results above require $\rho > 1$. In the limit $\alpha \searrow 0, \beta \searrow 0, \rho \searrow 1$ the η^{-1}, ζ^{-1} weights blow up. We rescale so that logarithms of edge weights converge to exponential random variables and bulk weights vanish. Let $\varepsilon > 0, \rho = \varepsilon\alpha + \varepsilon\beta + 1$, and consider the weights $(\xi_{\mathbb{N}^2}^{(\varepsilon)}, \eta_{\mathbb{N} \times \mathbb{Z}_+}^{(\varepsilon)}, \zeta_{\mathbb{Z}_+ \times \mathbb{N}}^{(\varepsilon)})$ under the distribution $\mu^{\varepsilon\alpha, \varepsilon\beta}$. The independent weights of (8.1) now satisfy

$$(8.3) \quad \xi_{i,j}^{(\varepsilon)} \sim \text{Gamma}(\rho), \eta_{i,0}^{(\varepsilon)} \sim \text{Gamma}(\varepsilon\alpha), \zeta_{0,j}^{(\varepsilon)} \sim \text{Gamma}(\varepsilon\beta), \quad i, j \in \mathbb{N}.$$

We can construct the weights in (8.3) as functions of uniform variables as in (4.26). Then the following limits as $\varepsilon \searrow 0$ can be taken pointwise:

$$-\varepsilon \log \xi_{i,j}^{(\varepsilon)} \rightarrow 0, \quad -\varepsilon \log \eta_{i,0}^{(\varepsilon)} \rightarrow I_{i,0} \sim \text{Exp}(\alpha), \quad \text{and} \quad -\varepsilon \log \zeta_{0,j}^{(\varepsilon)} \rightarrow J_{0,j} \sim \text{Exp}(\beta).$$

The NE induction equations (2.12) converge to the equations

$$(8.4) \quad I_x = (I_{x-e_2} - J_{x-e_1})^+ \quad \text{and} \quad J_x = (J_{x-e_1} - I_{x-e_2})^+.$$

The RWRE transition probability converges to a deterministic transition:

$$\pi_{x,x+e_1}^{(\varepsilon)} = \frac{\eta_{x+e_1}^{(\varepsilon)}}{\eta_{x+e_1}^{(\varepsilon)} + \zeta_{x+e_2}^{(\varepsilon)}} \longrightarrow \mathbf{1}\{I_{x+e_1} < J_{x+e_2}\} \equiv \pi_{x,x+e_1}^{(0)} \quad \text{as } \varepsilon \searrow 0.$$

The limit leads to an invariant distribution for a last-passage system. Equations (8.4) describe inductively the increment variables

$$I_x = G_{0,x} - G_{0,x-e_1} \quad \text{and} \quad J_x = G_{0,x} - G_{0,x-e_2}$$

of a degenerate last-passage model with boundary weights $\{I_{i,0}, J_{0,j} : i, j \in \mathbb{N}\}$ and zero bulk weights. This distribution on $(I_{\mathbb{N} \times \mathbb{Z}_+}, J_{\mathbb{Z}_+ \times \mathbb{N}})$ is invariant for the environment seen from the location φ_n that starts at $\varphi_0 = 0$ and obeys the transition

$$(8.5) \quad \pi_{x,x+e_1}^{(0)} = \mathbf{1}\{I_{x+e_1} < J_{x+e_2}\} \quad \text{and} \quad \pi_{x,x+e_2}^{(0)} = \mathbf{1}\{I_{x+e_1} > J_{x+e_2}\}.$$

Given the environment, this defines a deterministic path φ_\cdot on \mathbb{Z}_+^2 . We recognize in (8.5) the jump rule of the competition interface (2.4).

The remainder of this section is taken by the proofs. To prove stationarity of the Markov chain it suffices to consider the partial environment $(\eta_{\mathbb{N}e_1}, \zeta_{\mathbb{N}e_2}, \xi_{\mathbb{N}^2})$ because the other variables of the state are functions of these. The notation here is that $\eta_{\mathbb{N}e_1} = \{\eta_{ie_1}\}_{i \in \mathbb{N}}$, and similarly for other cases. The next lemma proves everything in Proposition 8.1 and Theorem 8.2 except the ergodicity.

LEMMA 8.6. Fix $n \in \mathbb{N}$ and an admissible path $x_{0,n}$ with $x_0 = 0$. Fix a finite set $\mathcal{I} \subset \mathbb{Z}_+^2$, disjoint from $(x_n + \mathbb{Z}_+^2) \cup \{x_k\}_{0 \leq k < n}$. Let $\{h_k\}_{k \in \mathbb{Z}_+}$ and $\{g_u\}_{u \in \mathbb{Z}_+^2}$ be collections of bounded Borel functions on \mathbb{R}_+ . Let f be a bounded Borel function on $\mathbb{R}_+^{\mathbb{N} + \mathbb{N} + \mathbb{N}^2}$. Then

$$(8.6) \quad \begin{aligned} & E^{\mu^{\alpha,\beta}} \left[P^\omega(X_{0,n} = x_{0,n}) \cdot \prod_{k=0}^{n-1} h_k(\check{\xi}_{x_k}) \cdot \prod_{u \in \mathcal{I}} g_u(\check{\xi}_u) \cdot f(\eta_{x_n + \mathbb{N}e_1}, \zeta_{x_n + \mathbb{N}e_2}, \xi_{x_n + \mathbb{N}^2}) \right] \\ &= P(X_{0,n} = x_{0,n}) \cdot \prod_{k=0}^{n-1} \mathbf{E}[h_k(G_{\alpha+\beta})] \cdot \prod_{u \in \mathcal{I}} \mathbf{E}[g_u(G_{\alpha+\beta+1})] \cdot E^{\mu^{\alpha,\beta}}[f(\eta_{\mathbb{N}e_1}, \zeta_{\mathbb{N}e_2}, \xi_{\mathbb{N}^2})]. \end{aligned}$$

Remark 8.7. Note that the independent $(\check{\xi}_{x_k})$ cannot go up to $k = n$ because $\check{\xi}_{x_n} = \eta_{x_n + e_1} + \zeta_{x_n + e_2}$ and these belong in the future of the walk. Adding the statements over $x_{0,n}$ gives the invariance of $\mu^{\alpha,\beta}$ and the distribution of $\check{\xi}$. For a fixed $x_{0,n}$ we get the averaged distribution of the walk and also the statement that when the walk looks at the $\check{\xi}$ weights in its past, it sees $G_{\alpha+\beta}$ -variables on its path and $G_{\alpha+\beta+1}$ -variables elsewhere.

Lemma 8.6 is basically a consequence of size-biasing beta variables. The formulation we need is in the next lemma, whose proof we leave to the reader.

LEMMA 8.8. Let the gamma variables below with distinct subscripts be independent. Then

$$(8.7) \quad \begin{aligned} & \mathbf{E} \left[\frac{G_\alpha}{G_\alpha + G_\beta} f \left(G_{\alpha+\beta+1} \cdot \frac{G_\alpha}{G_\alpha + G_\beta} \right) g \left(G_{\alpha+\beta+1} \cdot \frac{G_\beta}{G_\alpha + G_\beta} \right) h(G_\alpha + G_\beta) \right] \\ &= \frac{\alpha}{\alpha + \beta} \mathbf{E}f(G_{\alpha+1}) \cdot \mathbf{E}g(G_\beta) \cdot \mathbf{E}h(G_{\alpha+\beta}). \end{aligned}$$

Proof of Lemma 8.6. We assume that the first step of the walk is e_1 and calculate the distribution. Introduce functions Φ to represent north-east induction (2.12)–(2.13), specifically to calculate the $\check{\xi}$ weights on the vertical line $x \cdot e_1 = 0$ and ζ weights on the vertical line $x \cdot e_1 = 1$, for $x \cdot e_2 \geq 1$:

$$\begin{aligned} (\check{\xi}_{\mathbb{N}e_2}, \zeta_{e_1 + \mathbb{N}e_2}) &= (\check{\xi}_{\mathbb{N}e_2}, \zeta_{e_1 + e_2}, \zeta_{e_1 + e_2 + \mathbb{N}e_2}) \\ &= \left(\Phi_1(\eta_{e_1 + e_2}, \zeta_{e_2 + \mathbb{N}e_2}, \xi_{e_1 + e_2 + \mathbb{N}e_2}), \xi_{e_1 + e_2} \frac{\zeta_{e_2}}{\eta_{e_1} + \zeta_{e_2}}, \Phi_2(\eta_{e_1 + e_2}, \zeta_{e_2 + \mathbb{N}e_2}, \xi_{e_1 + e_2 + \mathbb{N}e_2}) \right). \end{aligned}$$

Let h_0, g, f_i be bounded Borel functions of their arguments. The first equality below implements definitions. In the second equality below apply (8.7) to the triple $(G_\alpha, G_\beta, G_{\alpha+\beta+1}) = (\eta_{e_1}, \zeta_{e_2}, \xi_{e_1 + e_2})$ and note that all other variables are independent of this triple. Let $G_{\alpha+\beta+1}^{\mathbb{N}e_2}$ denote an i.i.d. Gamma($\alpha + \beta + 1$) sequence. Augment temporarily the probability space with independent $G_{\alpha+1}$ and G_β variables that are also independent of all the other

variables in f_2 .

$$\begin{aligned}
& E^{\mu^{\alpha,\beta}} \left[P^\omega(X_1 = e_1) h_0(\check{\xi}_0) g(\check{\xi}_{\mathbb{N}e_2}) f_1(\eta_{e_1+\mathbb{N}e_1}) f_2(\zeta_{e_1+\mathbb{N}e_2}) f_3(\xi_{e_1+\mathbb{N}^2}) \right] \\
&= E^{\mu^{\alpha,\beta}} \left[\frac{\eta_{e_1}}{\eta_{e_1} + \zeta_{e_2}} h_0(\eta_{e_1} + \zeta_{e_2}) f_1(\eta_{e_1+\mathbb{N}e_1}) f_3(\xi_{e_1+\mathbb{N}^2}) \right. \\
&\quad \times g \left(\Phi_1 \left(\xi_{e_1+e_2} \frac{\eta_{e_1}}{\eta_{e_1} + \zeta_{e_2}}, \zeta_{e_2+\mathbb{N}e_2}, \xi_{e_1+e_2+\mathbb{N}e_2} \right) \right) \\
&\quad \times f_2 \left(\xi_{e_1+e_2} \frac{\zeta_{e_2}}{\eta_{e_1} + \zeta_{e_2}}, \Phi_2 \left(\xi_{e_1+e_2} \frac{\eta_{e_1}}{\eta_{e_1} + \zeta_{e_2}}, \zeta_{e_2+\mathbb{N}e_2}, \xi_{e_1+e_2+\mathbb{N}e_2} \right) \right) \Big] \\
(8.8) \quad &= \frac{\alpha}{\alpha + \beta} \mathbf{E}[h_0(G_{\alpha+\beta})] E^{\mu^{\alpha,\beta}}[f_1(\eta_{e_1+\mathbb{N}e_1})] E^{\mu^{\alpha,\beta}}[f_3(\xi_{e_1+\mathbb{N}^2})] \\
&\quad \times E^{\mu^{\alpha,\beta}} \left[g \left(\Phi_1(G_{\alpha+1}, \zeta_{e_2+\mathbb{N}e_2}, \xi_{e_1+e_2+\mathbb{N}e_2}) \right) \right. \\
&\quad \times f_2(G_\beta, \Phi_2(G_{\alpha+1}, \zeta_{e_2+\mathbb{N}e_2}, \xi_{e_1+e_2+\mathbb{N}e_2})) \Big] \\
&= \frac{\alpha}{\alpha + \beta} \mathbf{E}[h_0(G_{\alpha+\beta})] \mathbf{E}[g(G_{\alpha+\beta+1}^{\mathbb{N}e_2})] E^{\mu^{\alpha,\beta}}[f_1(\eta_{\mathbb{N}e_1})] E^{\mu^{\alpha,\beta}}[f_2(\zeta_{\mathbb{N}e_2})] E^{\mu^{\alpha,\beta}}[f_3(\xi_{\mathbb{N}^2})] \\
&= \frac{\alpha}{\alpha + \beta} \mathbf{E}[h_0(G_{\alpha+\beta})] \mathbf{E}[g(G_{\alpha+\beta+1}^{\mathbb{N}e_2})] E^{\mu^{\alpha,\beta}}[f_1(\eta_{\mathbb{N}e_1}) f_2(\zeta_{\mathbb{N}e_2}) f_3(\xi_{\mathbb{N}^2})].
\end{aligned}$$

In the second last equality, inside f_1 and f_3 we simply shift by $-e_1$. Inside f_2 variable G_β furnishes ζ_{e_2} . Here is the key point: at this stage the Burke property applies to the mappings (Φ_1, Φ_2) because $G_{\alpha+1}$ furnishes $\eta_{e_1+e_2}$ and thereby the parameters of the input weights satisfy $(\alpha + 1) + \beta = \rho$. The beta size-biasing put us back into the setting of a gamma system. Thus (Φ_1, Φ_2) outputs two independent sequences. The first one denoted by $G_{\alpha+\beta+1}^{\mathbb{N}e_2}$ is i.i.d. Gamma($\alpha + \beta + 1$) and it represents the distribution of $\check{\xi}_{\mathbb{N}e_2}$. The second one is i.i.d. Gamma(β), which we take to be $\zeta_{e_2+\mathbb{N}e_2}$. In the last equality we can combine the three $\mu^{\alpha,\beta}$ -expectations because the independence is in accordance with the definition of $\mu^{\alpha,\beta}$.

Standard arguments generalize the product $f_1 f_2 f_3$ so that

$$\begin{aligned}
(8.9) \quad & E^{\mu^{\alpha,\beta}} \left[P^\omega(X_1 = e_1) h_0(\check{\xi}_0) g(\check{\xi}_{\mathbb{N}e_2}) F(\eta_{e_1+\mathbb{N}e_1}, \zeta_{e_1+\mathbb{N}e_2}, \xi_{e_1+\mathbb{N}^2}) \right] \\
&= p(e_1) \mathbf{E}[h_0(G_{\alpha+\beta})] \mathbf{E}[g(G_{\alpha+\beta+1}^{\mathbb{N}e_2})] E^{\mu^{\alpha,\beta}}[F(\eta_{\mathbb{N}e_1}, \zeta_{\mathbb{N}e_2}, \xi_{\mathbb{N}^2})].
\end{aligned}$$

for Borel functions h_0, g, F such that the expectations make sense. Reflection across the diagonal gives the alternative formula where the first step is e_2 instead of e_1 , $\check{\xi}_{\mathbb{N}e_2}$ is replaced by $\check{\xi}_{\mathbb{N}e_1}$, and $G_{\alpha+\beta+1}^{\mathbb{N}e_2}$ is replaced by $G_{\alpha+\beta+1}^{\mathbb{N}e_1}$.

Referring to the goal (8.6), let $\mathcal{I}_0 = \mathcal{I} \setminus (x_1 + \mathbb{Z}_+^2)$ and take $g(\check{\xi}_.) = \prod_{u \in \mathcal{I}_0} g_u(\check{\xi}_u)$. We can combine the e_1 and e_2 cases into this statement, which is (8.6) for $n = 1$.

$$\begin{aligned}
(8.10) \quad & E^{\mu^{\alpha,\beta}} \left[P^\omega(X_1 = x_1) h_0(\check{\xi}_0) \cdot \prod_{u \in \mathcal{I}_0} g_u(\check{\xi}_u) \cdot F(\eta_{x_1+\mathbb{N}e_1}, \zeta_{x_1+\mathbb{N}e_2}, \xi_{x_1+\mathbb{N}^2}) \right] \\
&= p(x_1) \mathbf{E}[h_0(G_{\alpha+\beta})] \cdot \prod_{u \in \mathcal{I}_0} \mathbf{E}[g_u(G_{\alpha+\beta+1})] \cdot E^{\mu^{\alpha,\beta}}[F(\eta_{\mathbb{N}e_1}, \zeta_{\mathbb{N}e_2}, \xi_{\mathbb{N}^2})].
\end{aligned}$$

To obtain (8.6), do induction on the length n of the path. Let $\mathcal{I}' = \mathcal{I} \cap (x_1 + \mathbb{Z}_+^2)$. In (8.10) take

$$\begin{aligned} F(\eta_{\mathbb{N}e_1}, \zeta_{\mathbb{N}e_2}, \xi_{\mathbb{N}^2}) &= \prod_{i=1}^{n-1} \pi_{x_i - x_1, x_{i+1} - x_1}(\omega) \cdot \prod_{k=1}^{n-1} h_k(\check{\xi}_{x_k - x_1}) \cdot \prod_{u \in \mathcal{I}' - x_1} g_{u+x_1}(\check{\xi}_u) \\ &\quad \cdot f(\eta_{x_n - x_1 + \mathbb{N}e_1}, \zeta_{x_n - x_1 + \mathbb{N}e_2}, \xi_{x_n - x_1 + \mathbb{N}^2}). \end{aligned}$$

Assuming (8.6) holds for paths of length $n-1$, the right-hand side of (8.10) turns into the right-hand side of (8.6). \square

The ergodicity claim of Theorem 8.2 is in the next lemma.

LEMMA 8.9. *With initial distribution $\mu^{\alpha, \beta}$, the stationary process $S_n = (\eta_{X_n + \mathbb{N}e_1}, \zeta_{X_n + \mathbb{N}e_2}, \xi_{X_n + \mathbb{N}^2})$ is ergodic.*

Proof. Denote a generic state by $S = (\eta_{\mathbb{N}e_1}, \zeta_{\mathbb{N}e_2}, \xi_{\mathbb{N}^2})$. It suffices to show that, for any function $f \in L^1(\mu^{\alpha, \beta})$, the averages

$$n^{-1} \sum_{k=0}^{n-1} E^S[f(S_k)]$$

converge to a constant in $L^1(\mu^{\alpha, \beta})$ [26, p. 91–95]. By approximation in $L^1(\mu^{\alpha, \beta})$, it suffices to prove this for a local function f , that is, a function of the variables $\mathbf{s} = (\eta_{i,0}, \zeta_{0,j}, \xi_{i,j})_{i,j \in [M]}$ for an arbitrary but fixed $M \in \mathbb{N}$. Let $\mathbf{s} = \varphi(S)$ denote the projection mapping, and let the projection of the stationary process S_n be $\mathbf{s}_n = \varphi(S_n) = (\eta_{X_n + (i,0)}, \zeta_{X_n + (0,j)}, \xi_{X_n + (i,j)})_{i,j \in [M]}$.

Process \mathbf{s}_n is also a stationary Markov chain, with state space $\mathbb{R}_+^{2M+M^2}$ and invariant distribution $\nu = \mu^{\alpha, \beta} \circ \varphi^{-1}$. Under ν coordinates of \mathbf{s} are independent with distributions $\eta_{i,0} \sim \text{Gamma}(\alpha)$, $\zeta_{0,j} \sim \text{Gamma}(\beta)$ and $\xi_{i,j} \sim \text{Gamma}(\rho)$.

Given the state $\mathbf{s} = (\eta_{i,0}, \zeta_{0,j}, \xi_{i,j})_{i,j \in [M]}$, we compute the variables $\{\eta_x, \zeta_x : x \in [M]^2\}$ via north-east induction (2.12). The transition from state \mathbf{s} to a new state goes by two steps: (i) randomly shift \mathbf{s} by e_1 or e_2 ; (ii) add fresh variables to the north or east to replace the variables lost from south or west in the shift of the $M \times M$ square.

Precisely speaking, from $\mathbf{s} = (\eta_{i,0}, \zeta_{0,j}, \xi_{i,j})_{i,j \in [M]}$ the process jumps to either \mathbf{t}' or \mathbf{t}'' , according to the following two cases.

- (a) The shift is e_1 and $\mathbf{t}' = (\eta_{i+1,0}, \zeta_{1,j}, \xi_{i+1,j})_{i,j \in [M]}$ where the new independently chosen variables are $\eta_{M+1,0} \sim \text{Gamma}(\alpha)$ and $\xi_{M+1,j} \sim \text{Gamma}(\rho)$ for $j \in [M]$.
- (b) The shift is e_2 and $\mathbf{t}'' = (\eta_{i,1}, \zeta_{0,j+1}, \xi_{i,j+1})_{i,j \in [M]}$ where the new independently chosen variables are $\zeta_{0,M+1} \sim \text{Gamma}(\beta)$ and $\xi_{i,M+1} \sim \text{Gamma}(\rho)$ for $i \in [M]$.

The probabilities of the two alternatives are

$$\pi(\mathbf{s}, \mathbf{t}') = \frac{\eta_{e_1}}{\eta_{e_1} + \zeta_{e_2}} \quad \text{and} \quad \pi(\mathbf{s}, \mathbf{t}'') = \frac{\zeta_{e_2}}{\eta_{e_1} + \zeta_{e_2}}.$$

Let $\pi(\mathbf{s}, d\mathbf{t})$ denote the transition probability of the Markov chain \mathbf{s}_n : the shift followed by the random choice of new coordinates to complete the square $[M] \times [M]$. The task is to check that \mathbf{s}_n is an ergodic process.

Two general observations about checking the ergodicity of a Markov transition P with invariant distribution ν . (i) Suppose ν has a density with respect to a background measure λ . Then it is enough to check that, for ν -a.e. x , $P(x, dy)$ has a density $p(x, y)$ with respect to $\lambda(dy)$ such that $p(x, y) > 0$ for λ -a.e. y . For then, if A is a ν -a.s. invariant measurable set such that $\nu(A^c) > 0$, taking $x \in A^c$ in

$$\mathbf{1}_A(x) = P(x, A) = \int_A p(x, y) \lambda(dy)$$

shows that $\lambda(A) = 0$ and thereby $\nu(A) = 0$. (ii) It is enough to check the ergodicity of some power P^m .

We show that for $m = 2M + 1$, $\pi^m(\mathbf{s}, d\mathbf{t})$ has a Lebesgue almost everywhere positive density on $\mathbb{R}_+^{2M+M^2}$. Let B be a Borel subset of $\mathbb{R}_+^{2M+M^2}$. Write

$$(8.11) \quad T_x \mathbf{s} = (\eta_{x+(i,0)}, \zeta_{x+(0,j)}, \xi_{x+(i,j)} : i, j \in [M])$$

for the shifted configuration in the $M \times M$ square.

$$(8.12) \quad \begin{aligned} \pi^m(\mathbf{s}, B) &= \sum_{x \in \mathbb{Z}_+^2 : |x|_1 = m} E^{\mu^{\alpha, \beta}} [\mathbf{1}_B(T_x \mathbf{s}) P_0^\omega \{X_m = x\} \mid \varphi(S_0) = \mathbf{s}] \\ &= \sum_{x \in \mathbb{Z}_+^2 : |x|_1 = m} E^{\mu^{\alpha, \beta}} [E^{\mu^{\alpha, \beta}} \{ \mathbf{1}_B(T_x \mathbf{s}) \mid \mathcal{H}_x \} P_0^\omega \{X_m = x\} \mid \varphi(S_0) = \mathbf{s}]. \end{aligned}$$

On the first line above $E^{\mu^{\alpha, \beta}}$ represents the choices of fresh coordinates while the shifts are in the quenched probability $P_0^\omega \{X_m = x\}$. After that we conditioned on the σ -algebra (Figure 4)

$$(8.13) \quad \mathcal{H}_x = \sigma \{ \eta_{\mathbb{N}e_1}, \zeta_{\mathbb{N}e_2}, \xi_x, \{ \xi_{i,j} : i \leq x \cdot e_1 - 1 \text{ or } j \leq x \cdot e_2 - 1 \} \}.$$

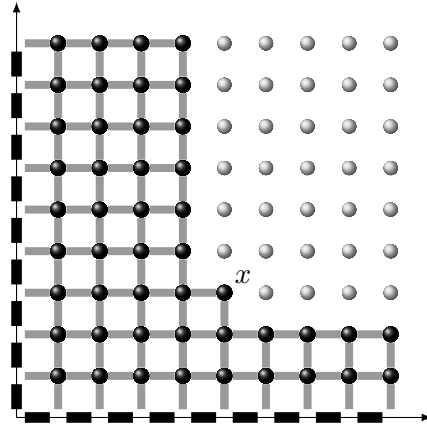


FIGURE 4. The σ -algebra \mathcal{H}_x . The dark black sites in the interior and the thickset lines on the axes denote the $\{\xi, \eta, \zeta\}$ variables that generate \mathcal{H}_x . The gray lines denote $\{\eta, \zeta\}$ variables computed via north-east induction from information contained in \mathcal{H}_x . Finally the lighter gray sites denote ξ variables independent of \mathcal{H}_x .

$X_m = x$ implies $|x|_1 = m$, and then $m = 2M + 1$ guarantees that \mathcal{H}_x is large enough to contain the event $\varphi(S_0) = \mathbf{s}$. The quenched probability $P_0^\omega\{X_m = x\}$ is also \mathcal{H}_x -measurable. Of the variables that make up $T_x \mathbf{s}$ in (8.11), the $\xi_{x+(i,j)}$'s are independent of \mathcal{H}_x , but the $\eta_{x+(i,0)}$'s and $\zeta_{x+(0,j)}$'s depend on \mathcal{H}_x through the equations

$$(8.14) \quad \eta_{x+(i,0)} = \xi_{x+(i,0)} \frac{\eta_{x+(i,-1)}}{\eta_{x+(i,-1)} + \zeta_{x+(i-1,0)}}, \quad i = 1, \dots, M$$

and

$$(8.15) \quad \zeta_{x+(0,j)} = \xi_{x+(0,j)} \frac{\zeta_{x+(-1,j)}}{\eta_{x+(0,j-1)} + \zeta_{x+(-1,j)}}, \quad j = 1, \dots, M.$$

The situations for $\{\eta_{x+(i,0)}\}$ and $\{\zeta_{x+(0,j)}\}$ are symmetric, so let us look at equation (8.14) closely. \mathcal{H}_x contains variables $\{\zeta_x; \eta_{x+(i,-1)} : i \in [M]\}$ because these can be computed by north-east induction from the variables listed in (8.13), so these are taken as given in (8.14). Variables $\{\xi_{x+(i,0)} : i \in [M]\}$ are picked i.i.d. Gamma(ρ), independently of \mathcal{H}_x , while variables $\{\zeta_{x+(i-1,0)} : i = 2, \dots, M\}$ are calculated along the way from the equations

$$(8.16) \quad \zeta_{x+(i-1,0)} = \xi_{x+(i-1,0)} \frac{\zeta_{x+(i-2,0)}}{\eta_{x+(i-1,-1)} + \zeta_{x+(i-2,0)}}, \quad i = 2, \dots, M.$$

Regarding $\{\zeta_x; \eta_{x+(i,-1)} : i \in [M]\}$ as given parameters, equations (8.14) and (8.16) show that the vectors $\bar{\eta} = (\eta_{x+(i,0)} : i \in [M])$ and $\bar{\xi} = (\xi_{x+(i,0)} : i \in [M])$ in $(0, \infty)^M$ are bijective functions of each other, and these functions are rational functions with positive coefficients. (The coefficients themselves are functions of $\{\zeta_x; \eta_{x+(i,-1)} : i \in [M]\}$.) Thus the Jacobians of these functions cannot vanish on $(0, \infty)^M$. Consequently, from the everywhere positive density of $\bar{\xi}$ (product of Gamma(ρ) distributions) we get an everywhere positive density f_1 for $\bar{\eta}$, for every given value of $\{\zeta_x; \eta_{x+(i,-1)} : i \in [M]\}$.

This argument can be repeated to get an everywhere positive density f_2 for the vector $\bar{\zeta} = (\zeta_{x+(0,j)} : j \in [M])$, for every given value of the variables specified by the conditioning on \mathcal{H}_x .

Let f denote the (everywhere positive) density of the vector $(\xi_{x+(i,j)} : i, j \in [M])$. With this notation we can write

$$E^{\mu^{\alpha,\beta}}[\mathbf{1}_B(T_x \mathbf{s}) | \mathcal{H}_x] = \int_{\mathbb{R}_+^{2M+M^2}} \mathbf{1}_B(u, v, w) f_1(u) f_2(v) f(w) du dv dw$$

where the right-hand side is not a constant, but the densities f_1 and f_2 depend also on the variables specified by the conditioning on \mathcal{H}_x . On the right the densities are multiplied due to independence that comes from dependence on disjoint sets of ξ variables. This formula can be substituted into (8.12) to conclude that $\pi^m(\mathbf{s}, \cdot)$ has an a.e. positive density on $(0, \infty)^{2M+M^2}$. \square

9. APPENDIX: AUXILIARY RESULTS

This section contains a comparison lemma for partition functions, a large deviation bound for the log-gamma polymer, and an ergodic theorem for correctors.

9.1. Comparison lemma for partition functions. Let arbitrary weights $\{V_x\}_{x \in \mathbb{Z}_+^2}$ be given and define partition functions as in (2.5). For a subset $A \subseteq \Pi_{u,v}$, define the restricted partition function (unnormalized polymer measure) by

$$(9.1) \quad Z_{u,v}(A) = \sum_{x \in A} \prod_{i=1}^{|v-u|_1} V_{x_i}^{-1}.$$

Recall the definitions of the exit points (3.8)–(3.9). The restriction $A = \{t_{e_1} > 0\}$ means that the first step of the path is e_1 . In other words, $Z_{0,x}(t_{e_1} > 0) = V_{e_1}^{-1} Z_{e_1,x}$, defined for $x \cdot e_1 \geq 1$.

LEMMA 9.1. *For $m \geq 2$ and $n \geq 1$ we have this comparison of partition functions:*

$$(9.2) \quad \frac{Z_{0,(m-1,n)}(t_{e_1} > 0)}{Z_{0,(m,n)}(t_{e_1} > 0)} \leq \frac{Z_{(1,1),(m-1,n)}}{Z_{(1,1),(m,n)}} \leq \frac{Z_{0,(m-1,n)}(t_{e_2} > 0)}{Z_{0,(m,n)}(t_{e_2} > 0)}.$$

Proof. Consider the ratio weights for these partition functions:

$$\begin{aligned} \eta_x &= \frac{Z_{0,x-e_1}(t_{e_1} > 0)}{Z_{0,x}(t_{e_1} > 0)} = \frac{Z_{e_1,x-e_1}}{Z_{e_1,x}} \quad \text{and} \quad \tilde{\eta}_x = \frac{Z_{(1,1),x-e_1}}{Z_{(1,1),x}}, \\ \zeta_x &= \frac{Z_{0,x-e_2}(t_{e_1} > 0)}{Z_{0,x}(t_{e_1} > 0)} = \frac{Z_{e_1,x-e_2}}{Z_{e_1,x}} \quad \text{and} \quad \tilde{\zeta}_x = \frac{Z_{(1,1),x-e_2}}{Z_{(1,1),x}}. \end{aligned}$$

On the boundary of the lattice \mathbb{N}^2 these ratios satisfy

$$\zeta_{1,j} = V_{1,j} = \tilde{\zeta}_{1,j} \quad \text{and} \quad \eta_{i,1} = V_{i,1} \frac{\eta_{i,0}}{\eta_{i,0} + \zeta_{i-1,1}} < V_{i,1} = \tilde{\eta}_{i,1} \quad \text{for } i, j \geq 2.$$

NE induction (2.12) preserves these inequalities, and gives the first inequality of (9.2). The second comes analogously. \square

9.2. Large deviation bound for the log-gamma polymer. Let $0 < \alpha < \rho$ and let (ξ, η, ζ) be a gamma system of weights with parameters (α, ρ) according to Definition 3.1. Let $Z_{0,v}$ be the partition function defined by (3.10) in this gamma system, with the corresponding point-to-point quenched polymer measure

$$(9.3) \quad Q_{0,v}\{x\} = \frac{1}{Z_{0,v}} \left(\prod_{i=1}^{t_{\text{exit}}} \tau_{\{x_{i-1}, x_i\}}^{-1} \right) \left(\prod_{j=t_{\text{exit}}+1}^{|v|_1} \xi_{x_j}^{-1} \right), \quad x \in \Pi_{0,v}.$$

Let the scaling parameter $N \geq 1$ be real valued. Let $(m, n) \in \mathbb{N}^2$ denote the endpoint of the path. Measure the deviation from characteristic velocity by

$$(9.4) \quad \kappa_N = |m - N\Psi_1(\rho - \alpha)| \vee |n - N\Psi_1(\alpha)|.$$

LEMMA 9.2. *Let κ_N be defined by (9.4). Let $\delta > 0$. Then there are constants $0 < \delta_1, c, c_1 < \infty$ such that the following estimate holds. For $(m, n) \in \mathbb{N}^2$, $N \geq 1$ and $u \geq (1 \vee c\kappa_N \vee \delta N)$,*

$$(9.5) \quad \mathbb{P}[Q_{0,(m,n)}\{t_{e_1} \geq u\} \geq e^{-\delta_1 u}] \leq e^{-c_1 u}.$$

Same bound holds for t_{e_2} . The same constants work for (α, ρ) that satisfy $0 < \alpha < \rho$ and vary in a compact set.

Proof. Let $\beta < \alpha$ and take two gamma systems: $(\xi, \eta^\alpha, \zeta^\alpha)$ with parameters (α, ρ) and $(\xi, \eta^\beta, \zeta^\beta)$ with parameters (β, ρ) . Couple them so that they share the ξ -variables, and $\eta_x^\beta \leq \eta_x^\alpha$ and $\zeta_x^\beta \geq \zeta_x^\alpha$ hold. This can be achieved by imposing these same conditions on the variables in part (c) of Definition 3.1, and then noting that the inequalities are preserved by (3.1). Let Z^α and Z^β be partition functions computed in these two systems.

$$(9.6) \quad \begin{aligned} Q_{0,(m,n)}\{t_{e_1} \geq u\} &= \frac{1}{Z_{0,(m,n)}^\alpha} \sum_{x \in \Pi_{0,(m,n)}} \mathbf{1}\{t_{e_1} \geq u\} \left(\prod_{i=1}^{t_{\text{exit}}} \frac{1}{\eta_{i,0}^\alpha} \right) \left(\prod_{j=t_{\text{exit}}+1}^{m+n} \xi_{x_j}^{-1} \right) \\ &\leq \frac{Z_{0,(m,n)}^\beta}{Z_{0,(m,n)}^\alpha} \cdot \prod_{i=1}^{\lfloor u \rfloor} \frac{\eta_{i,0}^\beta}{\eta_{i,0}^\alpha}. \end{aligned}$$

In the bounds below, $\bar{X} = X - \mathbb{E}X$ denotes a centered random variable. Recall the mean (3.11). Let $\delta_1 > 0$. From (9.6)

$$(9.7) \quad \begin{aligned} \mathbb{P}[Q_{0,(m,n)}\{t_{e_1} \geq u\} \geq e^{-\delta_1 u}] &\leq \mathbb{P}\left\{ \sum_{i=1}^{\lfloor u \rfloor} (\overline{\log \eta_{i,0}^\beta} - \overline{\log \eta_{i,0}^\alpha}) \geq \delta_1 u \right\} \\ &+ \mathbb{P}\left\{ \overline{\log Z_{0,(m,n)}^\beta} - \overline{\log Z_{0,(m,n)}^\alpha} \geq (\lfloor u \rfloor - m)(\Psi_0(\alpha) - \Psi_0(\beta)) \right. \\ &\quad \left. + n(\Psi_0(\rho - \beta) - \Psi_0(\rho - \alpha)) - 2\delta_1 u \right\}. \end{aligned}$$

Standard large deviations apply to log-gamma variables, so $\exists c_2 > 0$ such that

$$(9.8) \quad \mathbb{P}\left\{ \sum_{i=1}^{\lfloor u \rfloor} (\overline{\log \eta_{i,0}^\beta} - \overline{\log \eta_{i,0}^\alpha}) \geq \delta_1 u \right\} \leq e^{-c_2 u}.$$

Taylor expand to second order the Ψ_0 -differences inside the last probability in (9.7). Keeping $\delta > 0$ fixed, pick $\delta_1 > 0$ and $\alpha - \beta > 0$ small enough and $c < \infty$ large enough. Then for another small constant $c_3 > 0$, the probability simplifies to

$$(9.9) \quad \mathbb{P}\left\{ \overline{\log Z_{0,(m,n)}^\beta} - \overline{\log Z_{0,(m,n)}^\alpha} \geq c_3 u \right\} \leq e^{-c_4 u}.$$

The bound comes again from i.i.d. large deviations, by virtue of (3.12). \square

9.3. Ergodic theorem for correctors. With a bit of extra effort and with future use in mind we prove this ergodic theorem more generally than required for this paper. Fix a dimension $d \in \mathbb{N}$. Let $(\Omega, \mathfrak{S}, \mathbb{P})$ be a probability space equipped with a semigroup $(T_{x+y} = T_x \circ T_y)$ of measurable maps $T_x : \Omega \rightarrow \Omega$ for $x \in \mathbb{Z}_+^d$. Generic points of Ω are denoted by ω . Assume \mathbb{P} invariant and ergodic under $(T_x)_{x \in \mathbb{Z}_+^d}$: that is, $\mathbb{P} \circ T_x^{-1} = \mathbb{P}$, and if $T_x^{-1}A = A$ $\forall x \in \mathbb{Z}_+^d$ then $\mathbb{P}(A) \in \{0, 1\}$. Let $\mathcal{R} = \{e_i : i = 1, \dots, d\}$ denote the set of admissible steps, which are precisely the standard basis vectors of \mathbb{R}^d . Admissible paths $(x_k)_{k=0}^n$ satisfy $x_k - x_{k-1} \in \mathcal{R}$.

Let $F : \Omega \times \mathcal{R} \rightarrow \mathbb{R}$ be a corrector, by which we mean these properties:

$$(i) \quad \forall z \in \mathcal{R}: F(\omega, z) \in L^1(\mathbb{P}) \text{ and } \mathbb{E}F(\omega, z) = 0.$$

- (ii) The closed-loop (or cocyle) property: if $\{x_k\}_{k=0}^n$ and $\{x'_\ell\}_{\ell=0}^n$ are two admissible paths such that $x_0 = x'_0$ and $x_n = x'_n$, then

$$\sum_{k=0}^{n-1} F(T_{x_k} w, x_{k+1} - x_k) = \sum_{\ell=0}^{n-1} F(T_{x'_\ell} w, x'_{\ell+1} - x'_\ell).$$

Define the path integral of F by

$$f(x, \omega) = \sum_{i=0}^{n-1} F(T_{x_i} \omega, x_{i+1} - x_i), \quad (x, \omega) \in \mathbb{Z}_+^d \times \Omega,$$

where $(x_i)_{i=0}^n$ is any admissible path from $x_0 = 0$ to $x_n = x$. $f(0, \omega) = 0$. The closed-loop property ensures that f is well defined.

We make the following additional assumption.

$$(9.10) \quad \lim_{\delta \searrow 0} \overline{\lim}_{n \rightarrow \infty} \max_{x \in \mathbb{Z}_+^d : |x|_1 \leq n} \frac{1}{n} \sum_{0 \leq i \leq n\delta} |F(T_{x+iz} \omega, z)| = 0 \quad \forall z \in \mathcal{R}.$$

A similar assumption was useful in [24, 25] in studies of polymers. If, for each $z \in \mathcal{R}$, the variables $\{F(T_{iz} \omega, z)\}_{i \in \mathbb{Z}_+}$ are i.i.d., then by Lemma A.4 of [25] a sufficient condition for (9.10) is

$$(9.11) \quad \exists p > d : \mathbb{E}[|F(\omega, z)|^p] < \infty.$$

Our application of Theorem 9.3 is to the corrector $F^{\mathbf{u}}$ in (5.14). By Corollary 5.1 this satisfies the i.i.d. condition and even has an exponential moment. Thus hypothesis (9.10) is satisfied by $F^{\mathbf{u}}$ in (5.14).

THEOREM 9.3. *Let F be a corrector and satisfy (9.10). Then*

$$(9.12) \quad \lim_{n \rightarrow \infty} \max_{x \in \mathbb{Z}_+^d : |x|_1 = n} \frac{|f(x, \omega)|}{n} = 0.$$

As an auxiliary result towards the main theorem, we prove a limit for averages over rectangles of any dimension. The following result is a discrete version of Lemma 6.1 of [18].

THEOREM 9.4. *Let F be a corrector. Let $r \in [d]$, $0 \leq a_i < b_i$ for $1 \leq i \leq r$, and let*

$$(9.13) \quad B_{n,r} = \{x \in \mathbb{Z}_+^d : \lfloor na_i \rfloor \leq x_i < \lfloor nb_i \rfloor \text{ for } 1 \leq i \leq r, \ x_{r+1} = \dots = x_d = 0\}.$$

Then

$$(9.14) \quad \lim_{n \rightarrow \infty} \frac{1}{n^r} \sum_{x \in B_{n,r}} \frac{f(x, \omega)}{n} = 0 \quad \mathbb{P}\text{-a.s.}$$

It is enough to consider the case $a_i = 0$, for the general case is obtained by successive differences and sums of such cases. Then to simplify notation we take $b_i = 1$. We separate a part of the proof as a lemma.

LEMMA 9.5. Take $a_i = 0$ and $b_i = 1$ in (9.13). Let $1 \leq j \leq r \leq d$ and $g : [0, 1]^r \rightarrow \mathbb{R}$ continuous. Then

$$(9.15) \quad \lim_{n \rightarrow \infty} \frac{1}{n^r} \sum_{x \in B_{n,r}} g(n^{-1}(x_1, \dots, x_r)) F(T_x \omega, e_j) = 0 \quad \mathbb{P}\text{-a.s.}$$

Proof of Lemma 9.5. Fix j . The pointwise ergodic theorem [19, Thm. 6.2.8] gives

$$\lim_{n \rightarrow \infty} \frac{1}{n^r} \sum_{x \in B_{n,r}} F(T_x \omega, e_j) = h(\omega).$$

We wish to show that h is invariant under each shift T_{e_i} . By the closed-loop property (now for $j \in \{1, \dots, r\}$)

$$\begin{aligned} & \frac{1}{n^r} \sum_{k_1=0}^{n-1} \cdots \sum_{k_r=0}^{n-1} F(T_{k_1 e_1 + k_2 e_2 + \cdots + k_r e_r} \omega, e_j) \\ & \quad + \frac{1}{n^r} \sum_{k_1=0}^{n-1} \cdots \sum_{k_{j-1}=0}^{n-1} \sum_{k_{j+1}=0}^{n-1} \cdots \sum_{k_r=0}^{n-1} F(T_{k_1 e_1 + k_2 e_2 + \cdots + n e_j + \cdots + k_r e_r} \omega, e_i) \\ & = \frac{1}{n^r} \sum_{k_1=0}^{n-1} \cdots \sum_{k_{j-1}=0}^{n-1} \sum_{k_{j+1}=0}^{n-1} \cdots \sum_{k_r=0}^{n-1} F(T_{k_1 e_1 + k_2 e_2 + \cdots + 0 \cdot e_j + \cdots + k_r e_r} \omega, e_i) \\ & \quad + \frac{1}{n^r} \sum_{k_1=0}^{n-1} \cdots \sum_{k_r=0}^{n-1} F(T_{k_1 e_1 + k_2 e_2 + \cdots + k_r e_r} (T_{e_i} \omega), e_j). \end{aligned}$$

The closed loop above is undertaken for each fixed $T_{k_1 e_1 + \cdots + k_{j-1} e_{j-1} + k_{j+1} e_{j+1} + \cdots + k_r e_r} \omega$. The two paths are $\{e_j, 2e_j, \dots, ne_j, ne_j + e_i\}$ and $\{e_i, e_i + e_j, e_i + 2e_j, \dots, e_i + ne_j\}$.

The first sum converges to $h(\omega)$, the last one to $h(T_{e_i} \omega)$. By the pointwise ergodic theorem the first sum on the right converges to 0 because it has only n^{r-1} terms. Consequently all terms converge a.s. The second sum on the left must also vanish in the limit because it converges to zero in probability. We get $h(\omega) = h(T_{e_i} \omega)$, and conclude by ergodicity that $h = 0$. Then (9.15) follows by a Riemann sum-type approximation. \square

Proof of Theorem 9.4. This goes by induction on r . For $r = 1$, rearrange:

$$\frac{1}{n} \sum_{k=0}^{n-1} \frac{f(ke_1, \omega)}{n} = \frac{1}{n} \sum_{k=1}^{n-1} \frac{1}{n} \sum_{i=0}^{k-1} F(T_{ie_1} \omega, e_1) = \frac{1}{n} \sum_{k=0}^{n-1} \left(1 - \frac{k+1}{n}\right) F(T_{ke_1} \omega, e_1).$$

An application of (9.15) gives the conclusion (9.14) for $r = 1$.

Suppose that (9.14) holds for some $r \in \{1, \dots, d-1\}$. Let us show it for $r+1$.

$$\begin{aligned}
\frac{1}{n^{r+1}} \sum_{x \in B_{n,r+1}} \frac{f(x, \omega)}{n} &= \frac{1}{n^{r+2}} \sum_{x \in B_{n,r}} \sum_{k=0}^{n-1} f(x + ke_{r+1}, \omega) \\
&= \frac{1}{n^{r+2}} \sum_{x \in B_{n,r}} \sum_{k=0}^{n-1} [f(x, \omega) + f(ke_{r+1}, T_x \omega)] \\
&= \frac{1}{n^r} \sum_{x \in B_{n,r}} \frac{f(x, \omega)}{n} + \frac{1}{n^{r+2}} \sum_{x \in B_{n,r}} \sum_{k=0}^{n-1} f(ke_{r+1}, T_x \omega) \\
&= \frac{1}{n^r} \sum_{x \in B_{n,r}} \frac{f(x, \omega)}{n} + \frac{1}{n^{r+1}} \sum_{x \in B_{n,r+1}} \left(1 - \frac{x_{r+1} + 1}{n}\right) F(T_x \omega, e_{r+1}).
\end{aligned}$$

As $n \rightarrow \infty$ on the last line, the first sum goes to zero by the induction hypothesis and the second sum by (9.15). \square

Proof of Theorem 9.3. It is enough to prove

$$(9.16) \quad \lim_{n \rightarrow \infty} \min_{x \in \mathbb{Z}_+^d : |x|_1 = n} \frac{f(x, \omega)}{n} \geq 0.$$

This statement applied to both F and $-F$ verifies (9.12).

Let $\delta > 0$ and $a_k = k\delta/(4d)$ for $k \in \mathbb{Z}_+$. For $\mathbf{k} = (k_1, \dots, k_d) \in \mathbb{Z}_+^d$, define rectangles $B_{n,\mathbf{k}} = \{x \in \mathbb{Z}_+^d : \lfloor na_{k_i} \rfloor \leq x_i < \lfloor na_{k_i+1} \rfloor, i \in [d]\}$. For each $x \in \mathbb{Z}_+^d$ such that $|x|_1 = n$ we can pick a rectangle $B_{n,x} = B_{n,\mathbf{k}(x)}$ such that every point $y \in B_{n,x}$ can be reached from x with an admissible path of at most $n\delta$ steps. Our strategy is to replace $f(x, \omega)$ by an average of f over this rectangle. Note that there is a fixed finite set K of vectors \mathbf{k} such that the above choices can be made from $\{B_{n,\mathbf{k}} : \mathbf{k} \in K\}$ for all n and all $|x|_1 = n$.

For every x such that $|x|_1 = n$ and every $y \in B_{n,x}$ fix a path from x to y such that the steps e_1, e_2, \dots, e_d are taken in order. Then for any such pair x, y , with designated path $(x_i)_{i=0}^m$,

$$\begin{aligned}
f(x, \omega) &= f(y, \omega) - \sum_{i=0}^{m-1} F(T_{x_i} \omega, x_{i+1} - x_i) \\
&\geq f(y, \omega) - \sum_{z \in \mathcal{R}} \left\{ \max_{u \in \mathbb{Z}_+^d : |u|_1 \leq 2n} \sum_{0 \leq i \leq n\delta} |F(T_{u+iz} \omega, z)| \right\}.
\end{aligned}$$

The error term is independent of x, y . Average over $y \in B_{n,x}$ and then take minimum over $|x|_1 = n$:

$$\begin{aligned}
\min_{|x|_1 = n} \frac{f(x, \omega)}{n} &\geq \min_{\mathbf{k} \in K} \frac{1}{|B_{n,\mathbf{k}}|} \sum_{y \in B_{n,\mathbf{k}}} \frac{f(y, \omega)}{n} \\
&\quad - \sum_{z \in \mathcal{R}} \left\{ \max_{u \in \mathbb{Z}_+^d : |u|_1 \leq 2n} \frac{1}{n} \sum_{0 \leq i \leq n\delta} |F(T_{u+iz} \omega, z)| \right\}.
\end{aligned}$$

As $n \rightarrow \infty$, the first term on the right vanishes by Theorem 9.4. After that let $\delta \rightarrow 0$ and assumption (9.10) takes care of the last term. Bound (9.16) has been verified and Theorem 9.3 thereby proved. \square

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NICOS GEORGIOU, UNIVERSITY OF UTAH, MATHEMATICS DEPARTMENT, 155S 1400E, SALT LAKE CITY, UT 84112, USA.

E-mail address: georgiou@math.utah.edu

URL: <http://www.math.utah.edu/~georgiou>

FIRAS RASSOUL-AGHA, UNIVERSITY OF UTAH, MATHEMATICS DEPARTMENT, 155S 1400E, SALT LAKE CITY, UT 84112, USA.

E-mail address: firas@math.utah.edu

URL: <http://www.math.utah.edu/~firas>

TIMO SEPPÄLÄINEN, UNIVERSITY OF WISCONSIN-MADISON, MATHEMATICS DEPARTMENT, VAN VLECK HALL, 480 LINCOLN DR., MADISON WI 53706-1388, USA.

E-mail address: seppalai@math.wisc.edu

URL: <http://www.math.wisc.edu/~seppalai>

ATILLA YILMAZ, BOGAZICI UNIVERSITY, DEPARTMENT OF MATHEMATICS, 34342 BEBEK, ISTANBUL, TURKEY.

E-mail address: atilla.yilmaz@boun.edu.tr

URL: <http://www.math.boun.edu.tr/instructors/yilmaz>